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Sponsor Contact Person (s): Technical Matters  
Dr. M. S. Ghausi, Head  
Electrical Sciences and Analysis Section  
System Theory and Applications Program  
National Science Foundation  
Washington, D.C. 20550  
Phone: (202) 632-5881

Administrative Matters  
Mr. Wilbur W. Bolton, Jr.  
Grants Officer  
National Science Foundation  
Washington, D.C. 20550

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THE QUADRATIC SET COVERING (ASSIGNMENT)  
PROBLEM: APPLICATIONS AND COMPUTATION

Annual Report

June 1974

NSF Grant # GK - 38337

## THE QUADRATIC SET COVERING (ASSIGNMENT)

### PROBLEM: APPLICATIONS AND COMPUTATION

#### Summary

Linear set covering and set partitioning problems have attracted the attention of many researchers. This is due to the simple formulation of these problems and their wide ranges of applications. Even though the formulations of these problems is quite simple, solving them is challenging due to the difficulties involved with the 0-1 nature of the variables.

This research concerns itself with quadratic set covering and set partitioning problems. The significance of these problems is reflected by the flexibility of the quadratic objective, and its ability to deal with first order interaction between the variables, which leads to a wider range of applications. This report includes three sections. In the first section various potential applications of the quadratic set covering and quadratic assignment problems are presented. The second section deals with a branch and bound algorithm for obtaining optimal and suboptimal solutions of the quadratic assignment problem. In the last section we develop a branch and bound algorithm for solving the special case of the traveling salesman problem (symmetric and nonsymmetric). A pre-branch and bound dual optimization based on finding minimal 1-trees is used in order to speed the tree search.

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Section 1      The Quadratic Set Covering Problem,  
Related Problems and Applications

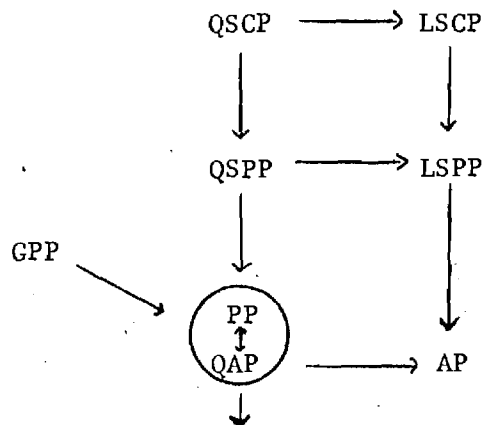
In this section we describe the quadratic set covering problem and eight other related problems, and give a number of examples illustrating the applications of these problems. Attention is given to the relationship between the problems, and a new generalized problem is presented with the threefold purpose of unifying several of the other problems, modelling more complicated extensions of the applications, and providing for more efficient solutions in some instances. Some examples are straightforward applications, while others serve to show how some applications are better modelled by quadratic than linear problems. One example points out the inadequacy of the quadratic set covering model, and indications are given for the development of a new problem. In some examples we point out the analogous factors in apparently diverse applications and identify complications unique to each application. Some new extensions of known applications are also given. [Garfinkel and Neurhanser (1972), Ch. 8]

1.1 Mathematical Descriptions of Nine Related Problems

We shall consider the following problems:

quadratic set covering problem	QSCP
quadratic set partitioning problem	QSPP
linear set covering problem	LSCP
linear set partitioning problem	LSPP
generalized placement problem (new)	GPP
placement problem	PP
quadratic assignment problem	QAP
assignment problem	AP
traveling salesman problem	TSP

All of the problems, except the GPP which is new, are well known. As the problems are precisely described below, it will be helpful to keep in mind the following diagram showing the relationships of the problems as special cases of the others in the formulations we give.



It should be noted that the mathematical formulations of the PP and QAP are identical, although the PP appears superficially to be more general. Hence, they are considered as a single problem here.

Let  $x$  be an  $n \times 1$  binary vector,  $x'$  the transpose of  $x$ ,  $A$  an  $m \times n$  matrix of zeros and ones,  $C$  an  $n \times n$  matrix, and  $e_m$  an  $m \times 1$  vector of ones. The quadratic set covering problem (QSCP) is

$$\begin{aligned} &\text{Minimize } x'Cx \\ &\text{such that } Ax \geq e_m \\ &x \text{ is a binary vector} \end{aligned}$$

and the quadratic set partitioning problem (QSPP) is

$$\begin{aligned} &\text{Minimize } x'Cx \\ &\text{such that } Ax = e_m \\ &x \text{ is a binary vector} \end{aligned}$$

If we let  $I_n$  be the  $n \times n$  identity matrix, then by choosing  $A$  to be the  $2n \times n^2$  matrix

$$A = \begin{bmatrix} e' & & & & \\ & e' & & & \\ & & \ddots & & \\ & & & e' & \\ I_n & I_n & \cdots & I_n & \end{bmatrix}$$

we obtain the quadratic assignment problem (QAP) for the case of assigning  $n$  objects to  $n$  locations. We may always assume, without loss of generality, that the number of objects to be assigned is equal to the number of locations, for if this is not the case, then we may add dummy objects which make them equal but have no effect on the final solution. By replacing the objective functions of the QSCP, QSPP, and QAP with a linear, instead of quadratic function, we obtain the linear set covering problem (LSCP), linear set partitioning problem (LSPP), and assignment problem (AP) respectively. The placement problem (PP) is simply the QAP with a linear expression  $d'x$  added to the objective function (here,  $d$  is an  $1 \times n$  cost vector), but the 0-1 variables  $x_i$  permit this linear expression to be absorbed into the quadratic portion of the objective function by adding the elements of  $d$  to the diagonal of  $C$ . Hence, from our viewpoint the PP and QAP are the same, although it may appear that the PP is more complicated when one is looking at real applications. The traveling

salesman problem (TSP), the problem of a salesman who must visit  $n$  cities, each exactly once, and return home while minimizing the total distance of his trip will be formulated later as a special case of the QAP. The generalized placement problem (GPP) includes the PP and QAP as special cases, but allows one to require that certain sets of objects must be assigned to certain sets of locations, or certain sets of locations must be filled by objects from certain sets of objects (Geddes 1974). As will be seen in the applications, these are very realistic restrictions, and when the GPP is formulated in terms of 0-1 variables the complications are actually helpful in obtaining the solution more efficiently. Before describing the GPP we need to introduce some notation. For a finite set  $S$  let  $|S|$  = the number of elements of  $S$  and  $P(S) = \{X | X \subset S\}$  = the power set of  $S$ . Suppose we are given two finite sets  $A$  and  $B$  with  $|A| = |B|$ . (Since  $A$  is thought of as a set of elements to be assigned to locations  $B$ , we must have  $|A| \leq |B|$ . There is no loss generality in assuming  $|A| = |B|$ , since "dummy" elements may always be added to  $A$  which do not influence the objective function.)  $\Phi = \{\psi | \psi: A \rightarrow B, \psi \text{ 1-1 map}\}$ . Assume that fixed assignment costs of elements of  $A$  to  $B$ , quantities shipped, and costs of shipping have been taken into consideration to produce a quadratic objective function, as in the QAP, and let  $c_\psi$  be the cost associated with  $\psi \in \Phi$ . Let  $R \subset P(A) \times P(B)$ . We say that an element  $\psi \in \Phi$  satisfies  $R$ , written  $\psi SR$ , provided:

$$\begin{aligned} \text{for each } (\hat{A}, \hat{B}) \in R, \quad & \psi[\hat{A}] \subset \hat{B} \quad \text{if } |\hat{A}| < |\hat{B}| \\ & \psi^{-1}[\hat{B}] \subset \hat{A} \quad \text{if } |\hat{A}| > |\hat{B}| \\ & \psi[\hat{A}] = \hat{B} \quad \text{if } |\hat{A}| = |\hat{B}| \end{aligned}$$

Here  $\psi[\hat{A}]$  denotes the image of the set  $\hat{A}$  under the map  $\psi$ . The generalized placement problem (GPP) is

$$\begin{aligned} & \text{minimize} \quad c_\psi \\ & \text{subject to} \quad \psi \in \Phi \\ & \text{and} \quad \psi SR \end{aligned}$$

where  $A$ ,  $B$ ,  $R$ , and the costs  $c_\psi$  are given. Note that if  $R = \{(A, B)\}$  we simply have the QAP. Other cases, such as when  $R$  is of the form  $R = \{(\{a_{i_1}\}, \{b_{i_1}\}), \dots, (\{a_{i_s}\}, \{b_{i_s}\})\}$ ,  $s < |A|$ , will be considered in sections 1.2 and 1.32, where examples are given for location problems and the TSP. A procedure is also given for writing the GPP in terms of 0-1 variables.



## 1.2 The Location Problems

Because of their similar mathematical structure as quadratic assignment problems, as well as analogous physical and economic interpretations, we give four applications under the general classification of location problems. These are

facilities location  
building layout  
control panel layout  
wiring problem

Each application will be discussed in detail individually, but first the location problem in general, including properties and extensions common to all of the applications, is described. We suppose that  $n$  objects  $a_i$  ( $i = 1, 2, \dots, n$ ) are to be assigned to  $n$  locations  $b_k$  ( $k = 1, 2, \dots, n$ ) in such a way that each object is assigned to one and only one location. As was mentioned earlier, it is easily seen that if we have  $m$  objects and  $n$  locations with  $m < n$ , we may supply  $n - m$  dummy objects so that there is no loss of generality in assuming the same number of objects and locations. There may or may not be a fixed cost  $f_{ik}$  associated with assigning object  $a_i$  to location  $b_k$ . There are given a fixed number of "units" to be "shipped" from object  $a_i$  to object  $a_j$ , say  $u_{ij}$ , and in general  $u_{ij} \neq u_{ji}$ . These units may be interpreted as trips made by people in the building layout problems, as we shall see, or as numbers of wire connections in the wiring problem. Shipping one unit from location  $b_k$  to  $b_\ell$  incurs a cost  $d_{k\ell}$  (usually related to distance). Define  $n^2$  variables  $x_{ik}$  ( $i = 1, 2, \dots, n$ ;  $k = 1, 2, \dots, n$ ) by

$$x_{ik} = \begin{cases} 1 & \text{if object } a_i \text{ is placed in location } b_k \\ 0 & \text{otherwise} \end{cases}$$

and let  $x' = (x_{1,1} \ x_{1,2} \ \dots \ x_{1,n} ; x_{n,1} \ x_{n,2} \ \dots \ x_{n,n})$ .

For  $i = 1, 2, \dots, n$  and  $j = 1, 2, \dots, n$ , define  $n^2$  matrices of order  $n \times n$

$$C_{ij} = \begin{pmatrix} c_{rs}^{ij} \end{pmatrix} \quad (r = 1, 2, \dots, n; \ s = 1, 2, \dots, n)$$

where, if  $i = j$ , then  $c_{rs}^{ij} = \begin{cases} f_{ir} & \text{if } r = s \\ 0 & \text{if } r \neq s \end{cases}$

$$\text{if } i \neq j, \text{ then } c_{rs}^{ij} = \begin{cases} 0 & \text{if } r = s \\ u_{ij} d_{rs} & \text{if } r \neq s \end{cases}$$

$c_{rs}^{ij}$  is therefore interpreted as the cost of shipping  $u_{ij}$  units from object  $a_i$  to object  $a_j$ , if they were located in  $b_r$  and  $b_s$  respectively  
 → Let  $C$  be the partitioned matrix

$$C = \begin{bmatrix} c_{11} & c_{12} & \dots & c_{1n} \\ c_{21} & c_{22} & \dots & c_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1} & c_{n2} & \dots & c_{nn} \end{bmatrix}$$

Let  $e_n$  be a  $n \times 1$  vector of ones and  $I_n$  be the  $n \times n$  identity matrix.

Let  $H$  be the  $2n \times n^2$  assignment matrix

$$H = \begin{bmatrix} e_n' & & & \\ & e_n' & & \\ & & \ddots & \\ & & & e_n' \\ I_n & I_n & \dots & I_n \end{bmatrix}$$

The location problem (QAP, PP) is to

$$\text{minimize } x' C x$$

$$\text{subject to } Hx = e_{2n}$$

$$x \text{ binary}$$

It should be noted that a large number of elements of the  $C$  matrix were set equal to zero since they correspond to assignments not permitted by the constraints  $Hx = e_{2n}$ . The first  $n$  rows of equations represented by  $Hx = e_{2n}$  require that each object is assigned to one location, and the remaining  $n$  equations require that each location is assigned to one object; that is to say we have a one-to-one correspondence between objects and locations. Now, for example, the object  $a_1$  cannot be assigned to locations  $b_1$  and  $b_3$  simultaneously, so the first row third column entry of

$c_{11}$ ,  $c_{13}$ , was set equal to zero. Actually, any number could be placed in this entry, since it would never contribute to the value of the objective functions  $x'Cx$  for any feasible  $x$ .

There are two extensions of the location problem which fall into the case of the GPP. In one of these we may require a few selected objects to be assigned to particular locations. For example, we may be given a collection of factories and warehouses already built and wish to add some new ones. A slightly different problem occurs when certain sets of objects must be assigned to certain sets of locations, or certain sets of locations must be filled by objects from given sets. In the second instance, we still have some freedom to locate the object. An example is that the tuning knob on a radio has to be, say, on the outside in front and not buried in the middle of the set, but there are several locations available. Another example is given below in which ten plants of various sizes, some of which require a water supply, are to be located on ten sites of different sizes, some of which have water supplies. This type of constraint may also be used to keep certain objects either close together or far apart. In a warehouse certain chemicals cannot be stored together for safety reasons such as fire and explosion. Certain areas in a warehouse may be unsuitable for storing some materials because of temperature or humidity, while some materials may be stored in any area of the warehouse. We may wish to require certain rooms in an office building to be located on an outside wall. In the design of electrical equipment some components may fail to function properly if located too far apart, while others fail when located too close together. An alternative to using the GPP to handle these complications would be to add constraints which simply require the distance between the locations containing objects be greater than (or less than) some fixed number. This may allow for lower objective function values, but may also give a difficult problem to solve.

### 1.21 Facilities Location

We begin by describing a simple example in which three plants  $a_i$  ( $i = 1, 2, 3$ ) are to be assigned to three locations  $b_k$  ( $k = 1, 2, 3$ ). The fixed costs  $f_{ik}$  of assigning plant  $a_i$  to location  $b_k$  are given in Table 1. Table 2 gives the number of units  $u_{ij}$  to be shipped from plant  $a_i$  to plant  $a_j$ , and Table 3 gives the unit costs  $d_{kl}$  of shipping from location  $b_k$  to  $b_l$ .

		Location $b_k$					Plant $a_j$		
		<div> <div><math>i \backslash k</math></div> <div>1 2 3</div> </div>					<div> <div><math>i \backslash k</math></div> <div>1 2 3</div> </div>		
Plant $a_i$	1	200	300	250	Plant $a_i$	1	-	50	100
	2	400	350	300		2	60	-	80
	3	600	700	650		3	40	70	-

Table 1. Values of  $f_{ik}$

Table 2. Values of  $u_{ij}$

		Location $b_l$		
		<div> <div><math>k \backslash l</math></div> <div>1 2 3</div> </div>		
Location $b_k$	1	-	2	3
	2	1	-	2
	3	3	3	-

Table 3. Values of  $d_{k\ell}$

We compute the C and H matrices described above:

$$C = \begin{bmatrix} c_{11} & c_{12} & c_{13} \\ c_{21} & c_{22} & c_{23} \\ c_{31} & c_{32} & c_{33} \end{bmatrix}$$

$$= \begin{bmatrix} 200 & - & - & - & 100 & 150 & - & 200 & 300 \\ - & 300 & - & 50 & - & 100 & 100 & - & 200 \\ - & - & 250 & 150 & 150 & - & 300 & 300 & - \\ - & 120 & 180 & 400 & - & - & - & 160 & 240 \\ 60 & - & 120 & - & 350 & - & 80 & - & 160 \\ 180 & 180 & - & - & - & 300 & 240 & 240 & - \\ - & 50 & 120 & - & 140 & 210 & 600 & - & - \\ 40 & - & 80 & 70 & - & 140 & - & 700 & - \\ 120 & 120 & - & 210 & 210 & - & - & - & 650 \end{bmatrix}$$

$$H = \begin{bmatrix} 1 & 1 & 1 & & & & & & \\ & & & 1 & 1 & 1 & & & \\ & & & & & & 1 & 1 & 1 \\ 1 & & & 1 & & & 1 & & \\ & 1 & & & 1 & & & 1 & \\ & & 1 & & & 1 & & & 1 \\ & & & 1 & & & 1 & & \\ & & & & 1 & & & & 1 \end{bmatrix}$$

As before, for  $i = 1, 2, 3$  ;  $k = 1, 2, 3$  ; let

$$x_{ik} = \begin{cases} 1 & \text{if plant } a_i \text{ is assigned to location } b_k \\ 0 & \text{otherwise} \end{cases}$$

$$x' = (x_{11} \ x_{12} \ x_{13} \ x_{21} \ x_{22} \ x_{23} \ x_{31} \ x_{32} \ x_{33})$$

$$\text{and } e'_6 = (1 \ 1 \ 1 \ 1 \ 1 \ 1).$$

The location problem is then

$$\begin{aligned} &\text{minimize} && x'Cx \\ &\text{subject to} && Hx = e_6 \\ &&& \text{and } x \text{ is a binary vector} \end{aligned}$$

It should be emphasized that in this problem the quantities shipped between plants,  $u_{ij}$ , were given and fixed, which in more general instances is not always the case. There are a number of complications which may be introduced. For example, the shipment of goods between plants, if placed in certain locations, may require the construction of a road or pipeline, and the fixed cost of such construction must be taken into consideration. It is also possible that more than one plant may be placed in one location. A very common problem is that we already have a number of plants built and wish to add a few more. Suppose we are considering  $n$  plants  $a_i$  and  $n$  locations  $b_k$ , but  $s < n$  of these plants are already built. In particular, assume that for  $j = 1, 2, \dots, s$  we have plant  $a_{i_j}$  built in location  $b_{i_j}$ . We can formulate this problem as a GPP by letting

$$R = \{(\{a_{i_1}\}, \{b_{i_1}\}), (\{a_{i_2}\}, \{b_{i_2}\}), \dots, (\{a_{i_s}\}, \{b_{i_s}\})\}$$

A second more complicated example of the GPP is given, and the relationship of the 0-1 variables shown. Let  $A = \{a_1, a_2, \dots, a_{10}\}$  be a set of ten plants and  $B = \{b_1, b_2, \dots, b_{10}\}$  be a set of ten locations. We are given that  $u_{ij}$  units are to be shipped from plant  $a_i$  to plant  $a_j$ , the cost of shipping one unit from location  $b_k$  to  $b_\ell$  is  $d_{k\ell}$ , and the cost of placing plant  $a_i$  in location  $b_k$  is  $f_{ik}$ .

$$\text{Let } A_1 = \{a_8, a_9, a_{10}\}$$

$$B_1 = \{b_8, b_9, b_{10}\}$$

$$A_2 = \{a_4, a_5, a_6, a_7, a_8\}$$

$$B_2 = \{b_1, b_2, b_3, b_4, b_5, b_8, b_9\}$$

$$A_3 = \{a_1, a_2, a_3, a_4, a_5, a_9\}$$

$$\text{and } B_3 = \{b_2, b_7, b_8\}$$

$$\text{Let } R = \{(A_1, B_1), (A_2, B_2), (A_3, B_3)\} \subset P(A) \times P(B)$$

We define  $C$  and  $H$  as before and obtain the problem:

$$\text{minimize } x'Cx \quad (1)$$

$$\text{subject to } Hx = e_{20} \quad (2)$$

$$x \text{ binary} \quad (3)$$

and elements from  $A_1$  must be placed in locations of  $B_1$ , and only those locations (4a)

elements from  $A_2$  must be placed in locations of  $B_2$  (4b)

elements placed in  $B_3$  locations must come from  $A_3$  (4c)

Statements (1)(2)(3) make up a QAP, and 4abc tell us that

$$x_{ij} = 0 \text{ for } i \text{ and } j \text{ such that } \begin{cases} a_i \in A_1 \text{ and } b_j \in B \setminus B_1 & (4a) \\ \text{or } a_i \in A \setminus A_1 \text{ and } b_j \in B_1 \\ a_i \in A_2 \text{ and } b_j \in B \setminus B_2 & (4b) \\ a_i \in A \setminus A_3 \text{ and } b_j \in B_3 & (4c) \end{cases}$$

(Here  $A \setminus A_1$  is the complement of  $A_1$  in  $A$ , etc.)

Interpretations of these constraints might be: plants of set  $A_2$  require a water supply which is available only in locations of set  $B_2$ ; locations of  $B_3$  are small and must be occupied by plants not requiring much space, those of  $A_3$ ; and plants of  $A_1$  must be built together in a certain region  $B_1$ .

### 1.22 Building Layout

The building layout problem is quite similar to the facilities location problem except that we are concerned with what goes on between various rooms in a building rather than shipments between plants. One objective may be to minimize the time employees spend walking between offices or rooms by finding optimal locations for the offices. For example, in the case of a hospital, we have a wide collection of employees such as surgeons, nurses, student nurses, etc., who collect various salaries and find it necessary (according to their job) to travel with varying degrees of frequency between different rooms such as an emergency room, operating room, scrub-up room, or sterilizing room. Now it has been pointed out that, on the average, people in a typical hospital spend 38 percent of their time walking between rooms, and taking relative salaries into account, this represents 34 percent of the cost for staff salaries. If we assume that careful planning can reduce this figure by one quarter, a savings equivalent to  $8\frac{1}{2}$  percent of the cost for staff salaries is realized. This may amount to a substantial sum of money, especially over a period of many years [Whitehead and Elders (1964)].

There are, however, some special problems in building layout which do not occur in any analogous manner in facilities location, control panel layout, or the wiring problem. We begin in a typical building layout problem by making a list of the rooms, and the frequency with which employees of different salaries travel between the rooms. For each pair of rooms  $a_i$  and  $a_j$  we may then compute a cost factor  $u_{ij}$  equal to the sum of the products of the number of people traveling between the two rooms and their salaries (say, in dollars/hours). The movements are then examined to see if they can be eliminated as unnecessary or new facilities or organization could lead to their elimination. For example, a recovery ward may be added so the surgeon or anaesthetist doesn't need to visit patients in their own room right after an operation. A set of location units  $b_k$  is determined. Since the rooms may vary in size, some may require only one location while others require several (adjacent locations. Instead of distance  $d_{k\ell}$

between locations  $b_k$  and  $b_\ell$ , we let  $d_{k\ell}$  represent the time (say, in fractions of hours) it takes to travel from  $b_k$  to  $b_\ell$ . Of course, a difficulty arises when a room occupies several locations, and in this case we may arbitrarily take the shortest (or longest) distance between the locations used by the rooms.

Once a placement has been determined, refinements may be needed such as the addition of a corridor to help maintain aseptic conditions in the operating room, or even changes for aesthetic purposes. It should be pointed out then, that the solution of the building layout problem as a QAP (or GPP) yields only an approximate solution to the real problem. Some factors such as the cost of air ducts, plumbing, and wiring between rooms we have neglected, because in most instances the differences are small for various layouts.

### 1.23 Control Panel Layout

One of the objectives in designing an instrument or control panel may be to minimize eye travel distance. Assume we are assigning  $n$  instruments  $a_i$  to  $n$  locations  $b_k$ . Using notation similar to before, we let  $u_{ij}$  be the frequency of transitions from instruments  $a_i$  to  $a_j$  and  $d_{k\ell}$  the Euclidean distance between locations  $b_k$  and  $b_\ell$ . As before, the cost and assignment matrices  $C$  and  $H$  may be computed to obtain a QAP:

$$\begin{array}{ll} \text{minimize} & x'Cx \\ \text{subject to} & Hx = e_{2n} \\ & x \text{ binary} \end{array}$$

This problem is discussed in [Dorris (1971), 18-21], while a linear case is given in [Freund and Sadosky (1967)].

### 1.24 The Wiring Problem

Beginning in the late 1950's, a problem associated with the construction of computers, now known as the backboard wiring problem, began to receive attention [Loberman and Weiberger (1957)], [Steinberg (1961)]. The problem is to find a placement (assignment) for a number of electrical components, which must be interconnected, on a board so that the total length of wire used to make the connections is minimal. An assumption is made that minimal total wire length is the best approximation for satisfying a number of goals for good circuit design, which include:



reducing capacitance  
 reducing delay line effects  
 distribution of heat sinks  
 avoidance of excessive wire build-up on routing channels  
 avoiding cross-talk without excessive use of expensive shielding  
 neatness to reduce construction and maintenance costs  
 using smallest amount of wire

We suppose a set of  $n$  electrical components  $a_i$  are given which must be assigned to a set of  $n$  locations  $b_k$ . As before, without loss of generality, it is assumed the number of components equals the number of locations, and they will be assigned in a one-to-one manner. Let  $u_{ij}$  be the number of wires connecting components  $a_i$  and  $a_j$ . The distance  $d_{kl}$  between  $b_k$  and  $b_l$  will, of course, depend on the wiring channels used and may be either Euclidean or "street" distance. The cost and assignment matrices may be found, as before, to give the problem:

$$\begin{aligned}
 &\text{minimize} && x' C x \\
 &\text{subject to} && H x = e_{2n} \\
 &&& x \text{ binary}
 \end{aligned}$$

Example: The best known example, and, in fact, about the only real problem which appears to have been studied in detail by many authors, is the one given in [Steinberg (1961)], [Graves and Whinston (1970)]. It is pointed out in [Hanan and Kurtzberg (1972)] that there is a need for studying additional realistic problems. Here we give a small example just to illustrate the ideas involved, the structure of the matrices, and some possible extensions. Consider the problem of placing four components  $a_i$  ( $i = 1, 2, 3, 4$ ), with  $u_{ij}$  wire connections between  $a_i$  and  $a_j$ , in four locations  $b_k$  ( $k = 1, 2, 3, 4$ ). The values of  $u_{ij}$  are given in Table 1 and the distances  $d_{kl}$  between locations  $b_k$  and  $b_l$  are given in Table 2. Figure 1 shows the actual arrangement of locations on the board. We have used Euclidean distances to compute the  $d_{kl}$ .

$i \backslash j$	1	2	3	4
1	0	3	7	1
2	3	0	2	5
3	7	2	0	3
4	1	5	3	0

Table 1. Values of  $u_{ij}$

$k \backslash l$	1	2	3	4
1	0	1	1	$\sqrt{2}$
2	1	0	$\sqrt{2}$	1
3	1	$\sqrt{2}$	0	1
4	$\sqrt{2}$	1	1	0

Table 2. Values of  $d_{kl}$  with Euclidean Distance

$b_1$	$b_2$
$b_3$	$b_4$

Figure 1

The assignment matrix H is

$$H = \begin{bmatrix} 1 & 1 & 1 & 1 & & & & & & & & & & & & \\ & & & & 1 & 1 & 1 & 1 & & & & & & & & \\ & & & & & & & & 1 & 1 & 1 & 1 & & & & \\ & & & & & & & & & & & & 1 & 1 & 1 & 1 \\ 1 & & & & 1 & & & & 1 & & & & 1 & & & \\ & 1 & & & & 1 & & & & 1 & & & & 1 & & \\ & & 1 & & & & 1 & & & & 1 & & & & 1 & \\ & & & 1 & & & & 1 & & & & 1 & & & & 1 \\ & & & & 1 & & & & 1 & & & & 1 & & & \\ & & & & & 1 & & & & 1 & & & & 1 & & \end{bmatrix}$$

The cost matrix C could be computed as before, but since the  $(u_{ij})$  and  $(d_{kl})$  matrices are symmetric, we may write C as the partitioned matrix

$$C = \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix}$$

$$\text{where } C_{ij} = \begin{pmatrix} u_{ij} \\ C_{rs} \end{pmatrix} \quad \begin{matrix} r = 1, 2, 3, 4, \\ s = 1, 2, 3, 4 \end{matrix}$$

$$\text{and } C_{rs}^{ij} = \begin{cases} u_{ij} d_{rs} & \text{if } i < j \quad \begin{matrix} (i = 1, 2, 3 \\ j = 2, 3, 4) \end{matrix} \\ & \text{and } r \neq s \\ 0 & , \quad \text{otherwise} \end{cases}$$

So C is the 16 x 16 matrix whose non-zero elements are shown:

C =

	0	3	3	$3\sqrt{2}$	0	7	7	$7\sqrt{2}$	0	1	1	$\sqrt{2}$
	3	0	$3\sqrt{2}$	3	7	0	$7\sqrt{2}$	7	1	0	$\sqrt{2}$	1
	3	$3\sqrt{2}$	0	3	7	$7\sqrt{2}$	0	7	1	$\sqrt{2}$	0	1
	$3\sqrt{2}$	3	3	0	$7\sqrt{2}$	7	7	0	$\sqrt{2}$	1	1	0
					0	2	2	$2\sqrt{2}$	0	5	5	$5\sqrt{2}$
					2	0	$2\sqrt{2}$	2	5	0	$5\sqrt{2}$	5
					2	$2\sqrt{2}$	0	2	5	$5\sqrt{2}$	0	5
					$2\sqrt{2}$	2	2	0	$5\sqrt{2}$	5	5	0
									0	3	3	$3\sqrt{2}$
									3	0	$3\sqrt{2}$	3
									3	$3\sqrt{2}$	0	3
									$3\sqrt{2}$	3	3	0

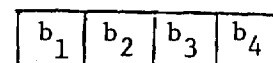
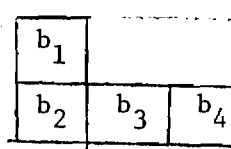
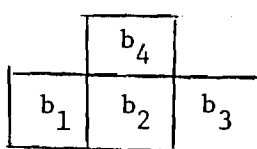
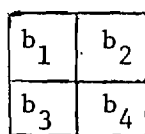
The problem is

$$\text{minimize } x' C x$$

$$\text{subject to } H x = e_8$$

x binary

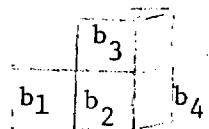
Some Extensions The above problem may be altered by introducing weights to the  $u_{ij}$ 's, as was done in the hospital example of the building layout problem, or by changing the  $(d_{kl})$  matrix. We may wish to change the  $(d_{kl})$  matrix to correspond to such physical arrangements of locations as



2 dimensional arrangements

1 dimensional

or possibly a 3 dimensional layout such as



If  $\mathcal{D}$  is a set of matrices

$$\mathcal{D} = \{ D = (d_{k\ell}) \}$$

we may wish to consider the problem of finding the best arrangement of locations as well:

$$\begin{array}{lll} \text{minimize} & \text{minimize} & x' C_D x \\ D \in \mathcal{D} & Hx = e & \\ & x \text{ binary} & \end{array}$$

(Here  $C_D$  indicates the cost matrix depends on  $D$ ).

As was done in the facilities location problem, we may use the GPP to require certain sets of components be located in specified sets of locations. This may be useful to keep the components of one section, say an amplification or power section, of a piece of electrical equipment together, as an aid to construction or maintenance.

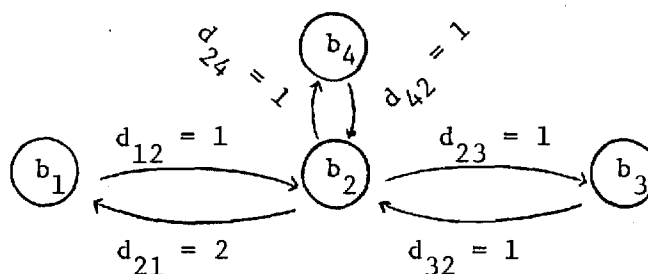
### 1.3 The Traveling Salesman Problem

Suppose a salesman is given a list of  $n$  cities connected by a number of roads of known distance. His problem, known as the traveling salesman problem, is to find a route of minimal length he can follow so as to visit each city once and only once, and return to where he started. (Hu (1969), 270-271], [Garfinkel and Neurhauser (1972), 354-361].

#### 1.31 The Classical Traveling Salesman Problem

There are several different versions and formulations of the TSP, and the one described above appears to be the original. We are concerned with stating the problem as a QAP, but first mention, for clarification, some of the variations. First, we are considering what is called the closed tour version. This simply means that the salesman must end up where he started. An alternative, is the open tour version in which the salesman starts in a given city, visits each city once and only once, but does not return to the original city. It is easy to convert the closed tour version to the open tour, by dividing the starting city into two cities. We are just going to consider the closed tour here.

Now the following example is a simple TSP which has no solution. We are given four cities  $b_k$  ( $k = 1, 2, 3, 4$ ) and some roads of known distances  $d_{kl}$  between them as shown below



It is easy to see that it is impossible to visit all four cities and not visit  $b_2$  three times. There is a modification of the TSP described in 1.32 which allows for the possibility of passing through cities, but as the classical TSP, this problem has no solutions. We also note that  $d_{12} = 1$  and  $d_{21} = 2$ , so that  $d_{12} \neq d_{21}$ . It is not necessary that the distances between two cities be the same each way. The idea of a subtour comes from formulations of the TSP in which variables  $x_{ij}$  are used:

$$x_{ij} = \begin{cases} 1 & \text{if the salesman travels from } b_i \text{ to } b_j \\ 0 & \text{otherwise} \end{cases}$$

Unless constraints are added to prevent it, a "solution" may occur consisting of more than one trip, as shown below:



The trips 1-2-3-1 and 4-5-6-7-4 are called subtours, and obviously cannot be a solution to any problem in which a salesman must begin and end his trip in the same city. A tour is a sequence for visiting each city exactly once and ending up in the starting city. In the formulation we now give for the TSP as a QAP, the problem of subtours does not arise.

Suppose we are given  $n$  cities  $b_k$  ( $k = 1, 2, \dots, n$ ) with distances  $d_{k\ell}$  from city  $b_k$  to  $b_\ell$ . We assume  $d_{k\ell} \geq 0$ , and in cases where there is no  $d_{k\ell}$  given (as when there is no direct road from  $b_k$  to  $b_\ell$ ) we assign some large number  $M$ . We think of the TSP as a facilities location problem in which  $n$  plants  $a_i$  ( $i = 1, 2, \dots, n$ ) are to be assigned to the  $n$  cities  $b_k$  ( $k = 1, 2, \dots, n$ ) under the conditions the units shipped from  $a_i$  to  $a_j$ ,  $u_{ij}$  are given by:

- i) for  $i = 1, 2, \dots, n-1$   $u_{i \ i+1} = 1$
- ii)  $u_{n \ 1} = 1$
- iii) all others  $u_{ij} = 0$

and we wish to minimize the total cost. Constructing cost and assignment matrices  $C$  and  $H$  as before, we have the TSP stated as the QAP

$$\begin{aligned} &\text{minimize} && x' C x \\ &\text{subject to} && Hx = e_{2n} \\ &&& x \text{ binary} \end{aligned}$$

A variable  $x_{ik}^* = 1$  in the final solution  $x^*$  means city  $k$  is visited as the  $i$ th city.

### 1.32 Four Modifications of the Traveling Salesman Problem

As noted in 1.31 there are some TSP's which have no solution because it is impossible to visit each city exactly once. There are other examples in which we could reduce the distance traveled if we could "pass through" a city, i.e., visit it more than once. Such an example is shown below in Figure 1.

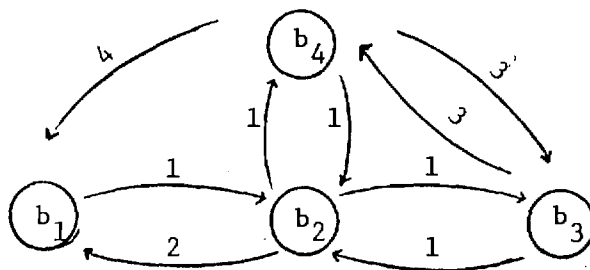
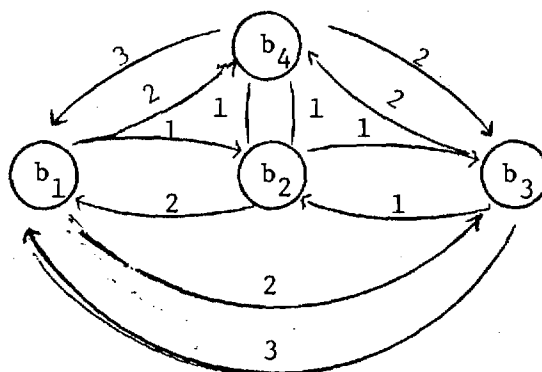


Figure 1

The shortest route to visit all cities by the TSP is 1-2-3-4-1 and has length 9. If we now permitted to visit  $b_2$  three times, we could choose the route 1-2-4-2-3-2-1 which has length 7. One method to avoid having the larger solution to the TSP forced upon us, which would be desirable in most real applications, is to construct an expanded TSP as follows. Given  $n$  cities and some values  $d_{kl}$  representing the distances of roads from city  $b_k$  to  $b_l$ , define values of  $\tilde{d}_{kl}$  for all  $k = 1, 2, \dots, n$ ;  $l = 1, 2, \dots, n$  by

$$\tilde{d}_{kl} = \begin{cases} \text{the shortest distance from} & \text{if } k \neq l \\ \text{city } b_k \text{ to } b_l & \\ \infty & \text{if } k = l \end{cases}$$

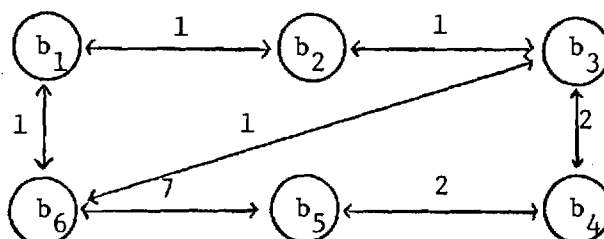
(We will need to keep track of the shortest paths when stating the solution of the original problem from the expanded problem). The example in Figure 1 would then become



with the optimal solution 1-2-4-3-1 and distance 7. If the original problem in Figure 1 had satisfied the triangle inequality for distances\*

$$d_{ij} + d_{jk} \geq d_{ik}$$

and we had been given values for all  $d_{k\ell}$  ( $k = 1, 2, \dots, n$ ;  $\ell = 1, 2, \dots, n$ ), then the optimal routing wouldn't require visiting  $b_2$  three times and the expanded TSP wouldn't have been needed [Hu (1969), 270]. The following example is one in which the given  $d_{k\ell}$  satisfies the triangle inequality, but a city is visited twice to obtain the shortest route 1-2-3-4-5-4-3-6-1 (length 12). The reason here is that not all the inter-distances among the cities are given.



We now mention some extensions of the TSP. In addition to being given  $n$  cities  $b_k$  ( $k = 1, 2, \dots, n$ ) and some distances  $d_{k\ell}$  between cities  $b_k$  and  $b_\ell$ , we may be given times  $t_{k\ell}$  between cities  $b_k$  and  $b_\ell$  and due dates  $\tau_k$  by which a city  $b_k$  must be visited. For a given tour  $\psi$ , let  $b_{\psi(i)}$  be the  $i$ th city visited, and suppose  $b_{\psi(i)}$  is visited at time

$$\hat{\tau}_{\psi(i)} = t_{\psi(0)\psi(1)} + t_{\psi(1)\psi(2)} + \dots + t_{\psi(i-1)\psi(i)}$$

for  $i = 1, 2, \dots, n$ . If we define the tardiness in visiting a city as

\*It doesn't:  $2 = 1 + 1 + d_{3,2} + d_{2,4} < d_{3,4} = 3$



$\hat{\tau}_k - \tau_k$ , one objective may be to find a tour  $\psi$  which minimizes the maximum tardiness. Another problem may be to minimize a linear combination of the distance traveled and the maximum tardiness. The GPP may be used to make such requirements as the first city visited must be one of three selected cities, or the last two cities visited must include a certain given city.

In a fourth modification of the TSP, considered by [Srivastava (1969)], the set of cities  $B = \{b_1, b_2, \dots, b_n\}$  is divided into a home city and a family of  $s$  collectively exhaustive and mutually exclusive sets of cities:  $B = \{\text{home city}\} \cup \bigcup_{i=1}^s B_i$ . The problem is to determine an optimal tour, starting at the home city, which includes at least one city from each of the sets  $B_i$  ( $i = 1, 2, \dots, s$ ). An application of this modification, in which a salesman has to complete multiple jobs subject to competition, is given in [Garg, etc. (1970)].

### 1.33 Production Scheduling Application

Suppose we have  $n$  tasks  $b_k$  ( $k = 1, 2, \dots, n$ ) which are to be performed on some machine. A change over time  $d_{k\ell}$  is required to change the machine from doing task  $b_k$  to task  $b_\ell$ . The problem of selecting a sequence in which to perform the tasks and return the machine to its original state so the sum of changeover times is minimized is a TSP [Maxwell (1964)].

An  $n$ -job,  $m$ -machine scheduling problem is modelled as a TSP in [Wisner (1972)].

### 1.34 The Multi-Traveling Salesmen Problem, Its Generalizations, and Some Applications

The TSP is actually a special case of the multi-traveling salesmen problem (M-TSP) [Svestka and Huckfeldt (1973)]. In the M-TSP we are given  $n$  cities  $b_k$  ( $k = 1, 2, \dots, n$ ), one of which is designated as a "home city", and distances  $d_{k\ell}$  between cities  $b_k$  and  $b_\ell$ . There are  $m$  salesmen instead of one, and the problem is to determine routes for each of the  $m$  salesmen to follow so that every city (except the home city) is visited exactly once by exactly one salesman, the home city is included in each salesman's route, and the total distance traveled by all salesmen is a minimum. We note that in the M-TSP a "home city" is designated beforehand. Three alternatives are to designate several home cities, as is done in the multi-terminal delivery problem [Tillman and Hering (1971)], or allow for either one or several undesignated home cities which will be determined as

part of the problem. (Not requiring a home city actually falls into the last two cases). Another generalization of the M-TSP is given in [Zak (1972)] where each salesman may have a different distance matrix.

We now give some applications of the M-TSP and its generalizations.

#### The Bank Messenger Scheduling Problem

In this problem a crew of messengers collects deposits at branch banks and returns them to a central office for processing. If there are  $m$  messengers and  $n$  branches, this is the M-TSP. The Cleveland Trust Company, Cleveland, Ohio, has actually made use of the M-TSP to determine routes for its messengers [Svestka and Huckfeldt (1973)].

#### Printing Press Scheduling for Multi-Edition Periodicals

A printing press scheduling problem is given as a 3-traveling salesmen problem with a number of side constraints in [Gorenstein (1970)]. But even a simplified version of this problem is not solvable for problem sizes with practical applications.

#### Scheduling Excavators Movements

The scheduling of the movements of excavators in mine pits is given in [Tsoi, etc. (1971)] as a M-TSP. Cases with route length and cycle constraints are given.

#### Single Terminal Delivery Problem

In this problem delivery trucks supply customers from a single terminal. This is a M-TSP, with the restriction that the trucks have limited capacity and need to return to the home city (terminal) to refill.

#### Multiterminal Delivery Problem

This is an application of the generalization of the M-TSP in which several home cities are designated. While in the single terminal delivery problem we assign customers to routes so total distance is minimized, in the multiterminal delivery problem we need to assign customers to routes and routes to terminals. This problem and references for the single terminal case may be found in [Tillman and Hering (1971)].

#### 1.4 Applications of the QSCP and Related Problems

In this section we give a number of examples of set covering problems, most of which were originally formulated as LSCP (or LSPP) but can be better formulated as QSCP (or QSPP). These include the delivery problem, location of fire hydrants, airline crew scheduling, assembly line balancing problem, information retrieval, project selection, capital budgeting, and attack/defense networks. A general structure common to all of these applications is first given, then each application is considered separately. The project selection problem is considered in detail and some directions are indicated for the formulation of a more realistic model.

We recall that the QSCP was

$$\text{minimize} \quad x'Cx \quad (1)$$

$$\text{such that} \quad Ax \geq b \quad (2)$$

$$x \text{ is a binary vector} \quad (3)$$

where  $x$ ,  $C$ ,  $A$ , and  $b$  were described in §1-1.  $A$  was an  $m \times n$  matrix. In the applications each of the  $m$  rows is identified with either a client (delivery problem), a flight segment (airline crew scheduling), a task (assembly line balancing), or some activity. Various subsets of the clients flight segments, tasks, or activities are selected to form (respectively) routes, trips, assembly stations, or projects, which are identified with the columns of  $A$ . For example, in the delivery problem, if client  $i$  is included in route  $j$ , then  $a_{ij} = 1$ , if not,  $a_{ij} = 0$ . The  $b$  vector ( $m \times 1$ ) has element  $b_i = 1$  in a particular problem if we wish the  $i$ th client to be included in the final solution. In the set covering problem,  $b_i = 0$  doesn't necessarily preclude the possibility of client  $i$  from being taken care of, as it does in the partitioning problem, but it doesn't require that client  $i$  be taken care of. If we don't care if a client is served several times we would use a covering problem; if we want to restrict a client to one service in a problem, the partitioning problem would be used. In the LSCP and LSPP, costs are assigned to the routes, trips, assembly stations, or projects (i.e. the columns of  $A$ ). The quadratic versions of these problems permit additional costs or discounts for various pairs of columns selected. In some applications there are good reasons why two activities should cost less than the sum of the costs of the activities

individually, while in other cases there appears to be no reason for extending the linear formulation to a quadratic one.

#### 1.41 Routing Problems: Truck Delivery and School Bussing

In the truck delivery and school bussing problems we are concerned with the delivery or pick-up of items, or, in the case of school bussing, children. There are a number of delivery or pick-up stations, each of which must be visited at least once, and a central terminal or school from which the items are sent out or collected. Various subsets of the delivery stations are formed to make up routes, and each route has a cost associated with it. The problem of selecting a set of routes so that each station is included in at least one route is a LSCP, or, if reasons can be found for quadratic interaction of the costs, a QSCP. In the case of school bussing one reason to use a QSCP might be to help achieve a racial balance. The question remains of how to do this. This would allow certain combinations of routes to be desirable and others to be undesirable. Since trucks or buses have limited capacity, the number of items picked up or delivered on a route is limited; also, since the total number of available trucks is limited (not allowing multiple trips for a vehicle), we may need a solution in which the total number of routes selected is less than or equal to the total number of vehicles [Balinski and Quandt (1964)], [Murty (1973)].

#### 1.42 Location of Fire Hydrants

The problem of locating fire hydrants is discussed in [Murty (1973)] We consider a network whose arcs are streets, and nodes are intersections of the streets. The problem is to select a subset of the nodes for placement of fire hydrants so each street has at least one fire hydrant. This problem is probably best left as a LSCP.

#### 1.43 Airline Crew Scheduling

In the airline crew scheduling problem flight segments are selected to form trips, and the idea is to select trips to cover all flight segments. This is a LSCP. See [Murty (1973)] and [Arabeyre, etc. (1969)].

#### 1.44 Assembly line balancing

Tasks are combined to be performed at assembly stations in the assembly line balancing problem. The problem is to determine which assembly stations are to be set up. This is a partitioning problem, since a task is to be

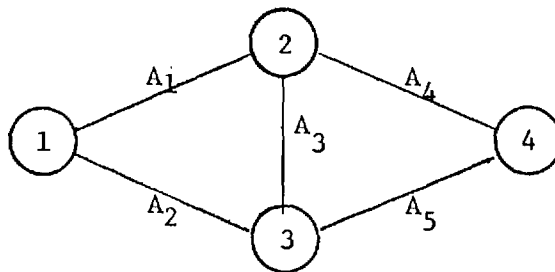
performed exactly once. Complications arise in the form of precedence relations on tasks and due dates by which tasks must be performed. [Murty (1973)], [Salvesan (1955)].

#### 1.45 Information Retrieval

The information retrieval problem is how to select a number of tapes from a collection of tapes, each of which contains subsets of required information, so that all desired information may be found searching a minimal length of tape. This is a LSCP and there appears to be no advantages in considering it as a QSCP. See [Day (1965)].

#### 1.46 Attack/Defense Networks

Attack and defense of networks may be formulated as QSCP. Consider the network shown below in Figure 1.



The nodes in the network are thought of as strategic points; while the arcs represent communication lines such as roads, bridges, pipelines, etc. The objective is to attack the network in an optimal manner so that nodes 1 and 4 are separated. If  $A_i$  represents arc  $i$  and  $x_i$  the associated variable, then  $x_i = 1$  means  $A_i$  is destroyed and  $x_i = 0$  means  $A_i$  is not destroyed. The constraint matrix which requires every path from node 1 to node 4 be cut at least once is

$$\begin{bmatrix}
 A_1 & A_2 & A_3 & A_4 & A_5 \\
 1 & 0 & 0 & 1 & 0 \\
 0 & 1 & 0 & 0 & 1 \\
 1 & 0 & 1 & 0 & 1 \\
 0 & 1 & 1 & 1 & 0
 \end{bmatrix}
 \begin{bmatrix}
 x_1 \\
 x_2 \\
 x_3 \\
 x_4 \\
 x_5
 \end{bmatrix}
 \geq
 \begin{bmatrix}
 1 \\
 1 \\
 1 \\
 1
 \end{bmatrix}$$

If a cost is assigned to destroying each arc, a linear objective function may be obtained to give a LSCP. However, the cost of destroying two arcs, say  $A_1$  and  $A_5$ , may be greater than the sum of the costs of destroying them individually. For example, they may be far apart and require some scarce resource to destroy. In this case a quadratic objective function  $x'Cx$  is used to consider such interactions. The element  $c_{15}$  would be positive and represent the additional cost of cutting  $A_1$  and  $A_5$ . The attack problem is a good example of a LSCP which is better modelled as a QSCP. The defense problem, how to optimally defend a network which will be attacked, by the construction of more arcs subject cut limitations, is also a QSCP. [Bellmore and Ratliff (1971a), (1971b)], [Jarvis (1968)].

#### 1.47 Project Selection and Capital Budgeting

We consider the problem where a number of projects are combined to form activities. A set of activities is to be selected in an optimal manner so that all required projects are included among the selected activities. Since it would probably be desired that a project is not included more than once, a partitioning problem would be used. Since there may be advantages of selecting certain pairs of activities, a QSCP is used.

To give some insight into the discount cost arising from the selection of two activities, consider the interaction which occurs in construction activities. There are three primary reasons why two activities are sometimes cheaper than the sum of the individual costs:

1. a materials discount When two activities contain projects which use the same type of material, a savings may be realized by buying this material in bulk. There is a limit to this savings, e.g. three \$5,000 projects may lead to a savings, but the addition of two more \$5,000 projects may not be significant,
2. already having equipment on the construction site To move a piece of machinery may cost \$400, \$100 to get a truck on the site and \$300 to move the machine. Perhaps three machines could be moved at the same time for  $100 + 3(300) = \$1000$  or a \$200 savings over  $3(400) = \$1200$ .
3. insurance savings (four jobs cost only a little more than one).

Unfortunately, there are also three major reasons why the QSPP doesn't

apply in many cases. First, we assume we know all of the projects (and activities) from the start, while in real problems new projects arise as work precedes on old ones. Then we assume all the projects are funded ahead of time and from one source. In practice, funding comes from several different sources, cannot be mixed, and is in various stages of uncertainty. Finally, as in construction work, there is a scheduling problem. We may need to hire two or three contractors at once, because one may not be able to simultaneously work on three projects which must be done by a certain date.

In practice then, a person with, say three, projects would probably

1. do none of them
2. do them all together, if possible
3. do separately as required by funding and scheduling.

Additional complications may occur in the project selection/capital budgeting problem in that

1. the selection of one activity may either necessitate or preclude the possibility of the selection of another activity
2. the total number of activities is limited
3. the total amount of money is limited (perhaps cancelling some optional activities).

It appears, in view of the difficulties described above, that a new model which takes the scheduling and funding, at least, into account may be developed. [Alexander (1973)] [Weingartner (1966)].

#### 1.48 Political Districting

In political districting an area (e.g. a state) is partitioned into a number of smaller areas called districts which are assigned a number of representatives. The area has also been partitioned into a number of basic population units (e.g. counties). A political districting plan consists in assigning the population units to the districts so that:

- i) a population unit is assigned to exactly one district
- ii) the total population assigned to a district doesn't deviate from the mean district population by more than a given number
- iii) a district is contiguous, i.e. it is possible to walk

between population units making up the district  
without going into another district (roughly  
speaking)

and iv) a district is compact, i.e. somewhat circular or square

Details and examples are given in [Garfinkel and Neurhauser (1970)].

Without the contiguity and compactness restrictions, this problem may

be formulated as a QSCP. Suppose we have  $n$  basic population units

$a_i$  ( $i = 1, 2, \dots, n$ ) with the population of unit  $a_i$  being  $d_i$ , and  $m$  districts

$b_k$  ( $k = 1, 2, \dots, m$ ). Define variables  $x_{ik}$  by

$$x_{ik} = \begin{cases} 1 & \text{if } a_i \text{ is assigned to } b_k \\ 0 & \text{otherwise} \end{cases}$$

and let  $x' = (x_{11} \ x_{12} \ \dots \ x_{1m} ; x_{21} \ x_{22} \ \dots \ x_{2m} ; \dots ; x_{n1} \ x_{n2} \ \dots \ x_{nm})$

The QSCP is then:

minimize  $x'Cx$

subject to  $Hx = e_n$

and  $x$  a binary vector

where  $C$  is the  $(n \cdot m) \times (n \cdot m)$  partitioned matrix

$$C = \begin{bmatrix} C_{11} & C_{12} & \dots & C_{1m} \\ \hline C_{21} & C_{22} & \dots & C_{2m} \\ \hline \vdots & & & \\ \hline C_{m1} & & & C_{mm} \end{bmatrix}$$

$$C_{ij} = \begin{bmatrix} d_1^2 - 2dd_1 & d_1 d_2 & \dots & d_1 d_n \\ d_1 d_2 & d_2^2 - 2dd_2 & & d_2 d_n \\ \vdots & \vdots & \ddots & \vdots \\ d_n d_1 & d_n d_2 & \dots & d_n^2 - 2dd_n \end{bmatrix} \quad \text{if } i = j$$



and  $C_{ij}$  = the  $n \times n$  zero matrix for  $i \neq j$

(here  $d = \frac{1}{m} \sum_{i=1}^n d_i$ , the mean district population)

and  $H$  is the  $n \times (n \cdot m)$  matrix

$$H = \begin{bmatrix} e_m' & & & \\ & e_m' & & \\ & & \ddots & \\ & & & e_m' \end{bmatrix}.$$

### 1.5 Miscellaneous Applications

The coloring problem [Bessiere (1965)], determining the minimum number of colors needed to color a map so that no two adjacent areas have the same color, is a LSPP. [Lawler (1960)] discusses the problem of minimizing "latency" in magnetic drum on disc storage computers as a QAP. The reader may see applications to staff and class scheduling, constrained least squares estimation, or message switching centers.

## 1.6 Some Comments on the Practicability of QSCP Applications

In this section we discuss the realistic values of the applications given in sections 1.2, 1.3, and 1.4. Quite often, simple examples are used to illustrate the application of some theory, and, while they serve the purpose of clarifying the theory, they are of no practical use.

Basically there are three reasons for this:

- i) the real problem is too simple to waste time applying a complicated theory
- ii) the theory is not complicated enough to model the real problem
- iii) the theory is OK for the real problem but leads to computationally infeasible problems.

We expect the first case (i) to hold for many facilities location and building layout problems. For example, a motel is a relatively uncomplicated structure and probably wouldn't require a QAP to determine a layout. The true value of the QAP is apparent when a complicated structure such as a hospital is designed or a complex wiring problem is considered. Unfortunately, in these cases the problem may become computationally infeasible (iii). One of the purposes of this study is to deal with such problems.

A good example where (ii) the theory was not complicated enough is the project selection problem of section 1.47. Directions were indicated here for a new problem.

Many of the references cited state that computationally infeasible problems result from the application of the theory described to real problems: [Jarvis (1968),88] states that although the attacker problem can be handled in a reasonably efficient manner, the defender problem cannot. Similar statements occur in [Maxwell (1964)] concerning production scheduling, in [Gorenstein (1970)] on printing press scheduling, and in [Freund and Sadosky (1967)] on control panel layout. In other words, there are a large number of valuable applications, once efficient solution procedures are developed. These procedures may include new algorithms and combinations/modifications of old ones, or the attempt to include more complications of the original problem as in the GPP.

How about the use of quadratic verse linear models for some applications? As was mentioned in section 1.4 some problems, e.g. the information retrieval problem, are naturally linear and no basis was found to consider quadratic

interaction. Examples of applications of set covering problems with quadratic objective functions, with the exception of the special case of the QAP, are not easy to find. Most examples considered, which were not linear, such as construction project selection, exhibit higher order interactions. It does not appear that examples with high order interactions can be approximated by quadratic interactions in a successful manner. Although [Weingartner (1966)] points out the tremendous advantage of the quadratic over the linear model for project selection/capital budgeting, some concrete, real applications remain to be found which can be solved as, say, a QSPP.

## REFERENCES

- Alexander, D., Interview, Physical Plant, Georgia Institute of Technology, 1973.
- Arabeyre, J. P., J. Fearnley, F. C. Steiger, and W. Teather, "The Airline Crew Scheduling Problem: A Survey", *Transportation Science* 3, 140-163, 1969.
- Balinski, M. L. and R. E. Quandt, "On an Integer Program for a Delivery Problem", *Operations Research* 13, 300-304, 1964.
- Bellmore, M. and H. D. Ratliff, "Set Covering and Involutory Bases", *Man. Sci.* 18, 194-205, 1971a.
- Bellmore, M. and H. D. Ratliff, "Optimal Defense of Multi-Commodity Networks", *Man. Sci.* 18, 174-185, 1971b.
- Bessiere, F., "Sur la Recherche der Nombre Chromatique d'Un Graph par uu Programme Lineaire en Nombres Entiers", *Rev. Franc, Recherche Operat.* 9, 143-148, 1965.
- Day, K. H., "On Optimal Extracting from a Multiple File Data Storage System: An Application of Integer Programming," *Operations Research* 13, 482-495, 1965.
- Dorris, A. L., "The Utility of Optimization Techniques in the Design of Man-Machine Systems", Master's Thesis, Georgia Institute of Technology, 1971.
- Freund, L. E. and T. L. Sadosky, "Linear Programming Applied to Optimization of Instrument Panel and Workplace Layout", *Human Factors* 9, 295-300, 1967.
- Garfinkel, K. S. and G. L. Nemhauser, "Optimal Political Districting by Implicit Enumeration Techniques", *Man. Sci.*, 16, 495-504, 1970b.
- Garfinkel, K. S. and G. L. Nemhauser, Integer Programming, John Wiley & Son, 1972.
- Garg, K. C., S. Kumar, P. Dass, and P. Len, "Generalized Traveling Salesman Problem Through n Sets of Nodes in a Competitive Market", *Ablanf und Plassungsforschung (Germany)* 11, 116-120, 1970.
- Geddes, P. H., "The Solution of Large Size Quadratic Set Covering Problems", Master's Thesis (in progress), Georgia Institute of Technology, 1974.

- Gorenstein, S., "Printing Press Scheduling for Multi-Edition Periodicals", *Man. Sci.* 16, 373-383, 1970b.
- Graves, G. W., and A. B. Whinston, "An Algorithm for the Quadratic Assignment Problem" in Integer and Nonlinear Programming, ed. by J. Abadie, North Holland Pub. Co., Amsterdam, 473-497, 1970.
- Hanan, M., and J. M. Kurtzberg (1972), "A Review of the Placement and Quadratic Assignment Problems", *SIAM Review* 14, 324-342, 1972.
- Hu, T. C., Integer Programming and Network Flows, Addison-Wesley, 1969.
- Jarvis, J. J., Optimal Attack and Defense of a Command and Control Communications Network, Ph.D. Dissertation, The Johns Hopkins University, 1968.
- Lawler, E. D., "Notes on the Quadratic Assignment Problem", Harvard Computation Laboratory, unpublished paper, 1960.
- Loberman, H., and A. Weinberger, "Formal Procedures for Connecting Terminals with a Minimum Total Wire Length", *Journal of the Association of Computing Machinery* 4, 428-437, 1957.
- Maxwell, W. L., "The Scheduling of Single Machine Systems: A Review", *The International Journal of Production Research* 3, 177-199, 1964.
- Murty, K. G., "On the Set Representation and Set Covering Problems", in Symposium on the Theory of Scheduling and its Applications, ed. by M. Beckman, G. Goos, and H. P. Kunzi, Springer-Verlag, 143-163, 1973.
- Salvesan, M. E., "The Assembly Line Balancing Problem", *Journal of Industrial Engineering* 6, 18-25, 1955.
- Srivastava, S. S., S. Kumar, K. C. Garg, and P. Len, "Generalized Traveling Salesman Problem through  $n$  Sets of Nodes", *CORS Journal* 7, 97-101, 1969.
- Steinberg, L., "The Backboard Wiring Problem: A Placement Algorithm", *SIAM Review* 3, 37-50, 1961.
- Svestka, J. A. and V. E. Huckfeld, "Computational Experience with an M-Salesman Traveling Salesman Algorithm", *Man. Sci.* 19, 790-799, 1973.
- Tillman, F. A., and R. W. Hering, "A Study of a Look-Ahead Procedure for Solving the Multiterminal Delivery Problem", *Transportation Research* 5, 225-229, 1971.

Tsoi, A. KH., S. M. Taskai, and V. N. Druzhinin, "The Scheduling of Movements of Excavators as a Problem of Several Traveling Salesmen", Trudy Instituta Gormogo Dela, Akademirijo Nauk Kazakhrkogo SSR (U.S.S.R.) 45, 9-13, 1971.

Weingartner, H. M., "Capital Budgeting of Interrelated Projects: Survey and Synthesis", Man Sci. 12, 458-516, 1966.

Whitehead, N. and M. Z. Elders, "An Approach to the Optimum Layout of Single-Storey Buildings", Architects' Journal 139, 1373-1380, 1964.

Wisner, D. A., "Solution of the Flowshop-Scheduling Problem with No Intermediate Queues", Operations Research 20, 689-697, 1972.

Zak, Yu. A., Solution Algorithms for "n Traveling Salesmen" Problem, Kibernetika (U.S.S.R.) 1, 99-106, (Russian, English Summary), 1972.

## Section II

### QUADRATIC ASSIGNMENT ALGORITHM

In this section, we will describe a branch and bound algorithm for solving the quadratic assignment problem. Recall that the problem is to assign  $m$  objects to  $n$  locations ( $n \geq m$ ). If object  $i$  is assigned to location  $\ell(i)$ , a fixed cost  $f_{i\ell(i)}$  is encountered. A flow or interaction  $u_{ij}$  between objects  $i$  and  $j$  results in a cost of the form  $u_{ij}d_{\ell(i)\ell(j)}$ , where  $d_{\ell(i)\ell(j)}$  is the distance between locations  $\ell(i)$  and  $\ell(j)$ . The total cost to be minimized can then be stated as follows.

$$\text{Minimize} \quad \sum_{i=1}^m f_{i\ell(i)} + \sum_{j=1}^m \sum_{\substack{i=1 \\ j \neq i}}^m u_{ij} d_{\ell(i)\ell(j)}$$

#### Decision Tree and General Framework

Since the problem is combinatorial in nature, all possible ways of assigning  $m$  objects to  $n$  locations, must be considered. Doing this explicitly, however, will result in a huge number of possibilities, and is impossible from a computational point of view. A large number of these possibilities must not be enumerated explicitly, if one is to develop a decent procedure. At each stage of the algorithm, we have a set of objects that has already been assigned to certain locations. These already assigned objects form a partial solution of the assignment problem. In order to obtain a feasible solution, i.e. a complete assignment, we must find a completion of the partial solution. Rather than considering all the possible ways of completing the partial solution, we first investigate whether the partial solution may lead to a complete solution, with an objective value which is smaller than the best complete solution that we already have. This is done by calculating a lower bound on the cost of the partial solution plus the cost of completing this partial solution.



Let  $B$  be the lower bound and let  $C^*$  be the cost of the best available assignment. If  $B \geq C^*$  then any completion of the partial solution can lead to no improvement. In this case, the partial solution is said to be fathomed, and is abandoned. On the other hand, if  $B < C^*$ , it is worthwhile to pursue the partial solution, by seeking to assign more objects.

#### Calculation of Lower Bounds On Cost of Completing a Partial Solution

Suppose that a set of objects in the set  $I$  have already been assigned to a set of locations  $J$ . In particular, suppose that object  $i$  is assigned to location  $\ell(i)$ . The cost of this partial solution,  $C_1$ , can be determined exactly as follows.

$$C_1 = \sum_{i \in I} f_i \ell(i) + \sum_{\substack{j \in I \\ j \neq i}} \sum_{i \in I} u_{ij}^d \ell(i) \ell(j)$$

Note that  $C_1$  consists of the fixed cost of assigning objects in the set  $I$  to their locations, plus the interaction among these objects.

The cost of completing this partial solution, i.e. the additional cost resulting from assigning the remaining objects, consist of the fixed costs, plus two types of interaction costs. The interaction from unassigned objects to assigned objects, with bound  $C_2$ , plus the interaction between the unassigned objects themselves, with bound  $C_3$ .

$C_2$ : bound on cost from unassigned objects to assigned objects.  $C_2$  is the optimal cost of the following linear assignment problem.

$$\begin{aligned} \text{Minimize} \quad & \sum_{j \notin J} \sum_{i \notin I} b_{ij} x_{ij} \\ \text{Subject to} \quad & \sum_{j \notin J} x_{ij} = 1 \quad \text{for each } i \notin I \\ & \sum_{i \notin I} x_{ij} \leq 1 \quad \text{for each } j \notin J \\ & x_{ij} \geq 0 \quad \text{for each } i \notin I, j \notin J \end{aligned}$$

where  $b_{ij}$  is a bound on the cost resulting from assigning the unassigned object  $i$  to the unassigned location  $j$ .

$$\text{e.g., } b_{ij} = f_{ij} + \sum_{t \in J} u_{it} d_{jl(t)} + \sum_{t \in J} u_{ti} d_{l(t)j}$$

In words,  $b_{ij}$  is the fixed cost of assigning object  $i$  to location  $j$ , plus the interaction from the assigned objects and object  $i$ , weighted by the distance from location  $j$  to the locations of the already assigned objects.

Of course, a complete solution of the linear assignment problem can be replaced by the simpler problem of reducing the matrix  $(b_{ij})$  such that it has a zero in each row and each column by subtracting the minima of the rows from the rows, and the minima of the columns from the resulting columns.  $C_2$  can then be replaced by the sum of the row minima plus the sum of the smallest  $m - |I|$  column minima, where  $|I|$  is the number of already assigned objects. Of course, this calculation is simpler, but the bound is not as tight as the bound obtained from solving the assignment problem.

$C_3$ : bound on cost between unassigned objects and themselves.

In order to calculate  $C_3$ , rank the interactions  $u_{ij}$  in a descending order for  $i, j \notin I$ , and rank the distances  $d_{ij}$  in an ascending order for  $i, j \notin J$ . This results in two vectors of size  $(m - |I|)^2$  and  $(n - |J|)^2$ . If  $n > m$ , complete the size of the first vector such that it has  $(n - |J|)^2$  entries by adding zeros to it from the bottom. Then  $C_3$  is the inner product of the two vectors. In words,  $C_3$  is calculated by matching the largest interaction among unassigned elements to the smallest distance between unassigned locations, and 2nd largest interaction to 2nd smallest distance, and so forth. Clearly, this procedure will give a lower bound on the cost among unassigned elements and themselves.

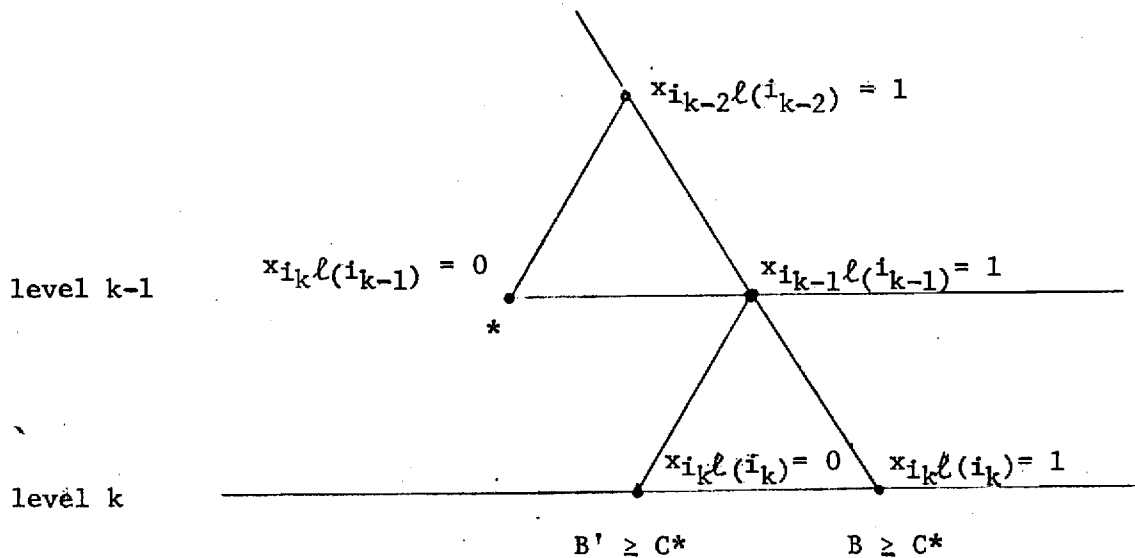
Now we can calculate the overall bound  $B$  as the cost of the partial solution plus all possible completions, namely,  $B = C_1 + C_2 + C_3$ .

## Continuation of the Search

### A. Fathoming (Backward Move)

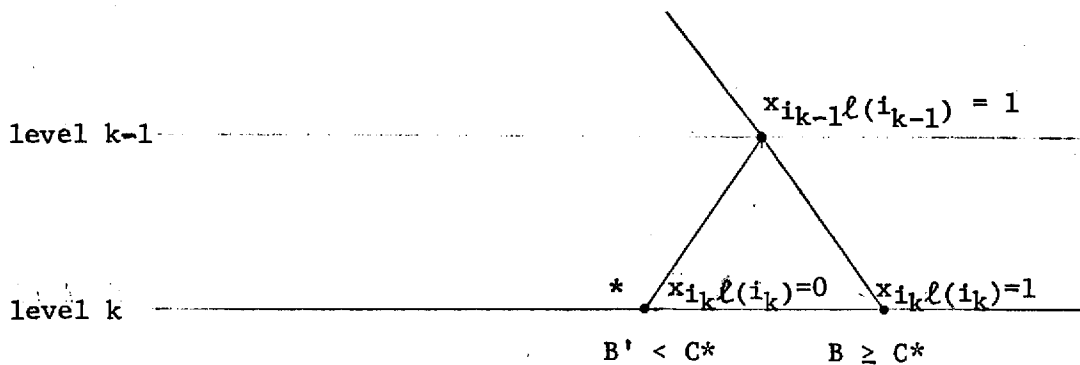
Suppose that  $k$  objects indexed by the set  $I$  have already been assigned to  $k$  locations indexed by the set  $J$ . The level of the tree is called  $k$ . A bound on the cost resulting from all completions of the current partial solution  $B$  is calculated as discussed earlier. If  $B \geq C^*$ , where  $C^*$  is the current best known cost of a complete assignment, then pursuing this partial solution is not worthwhile. The partial solution cannot lead to an improved complete solution, and is said to be fathomed. The last assignment, i.e. the  $k$ th assignment is banned, or prohibited, in the hope that this will lead to an improved completion. For example, if the  $k$ th assignment involved placing object  $i_k$  in location  $\ell_{(i_k)}$ , then  $x_{i_k \ell_{(i_k)}}$  is forced to be zero.  $i_k$  is then placed in the unassigned list of objects, (i.e.  $i_k$  removed from  $I$ ), and similarly  $\ell_{(i_k)}$  is removed from the unassigned list of locations (i.e.  $\ell_{(i_k)}$  is removed from  $J$ ). A new bound  $B'$  is calculated, in exactly the same manner as explained above, except of course, the assignment  $x_{i_k \ell_{(i_k)}} = 1$  is prohibited, say by forcing  $b_{i_k \ell_{(i_k)}} = +\infty$  while solving the linear assignment problem. If  $B'$  is still  $\geq C^*$ , then the partial solution of the first  $k-1$  assignments, while banning the assignment  $i_k$  to  $\ell_{(i_k)}$ , can still lead to no improved solutions. Since the first  $k-1$  assignments with  $x_{i_k \ell_{(i_k)}} = 1$  and  $x_{i_k \ell_{(i_k)}} = 0$  leads to no improvement, then all the possibilities at level  $k$  have been exhausted, and prohibiting the assignment at level  $k-1$  is now possible. This is called strong fathoming, and the level of the tree is thus reduced by 1 unit, by prohibiting the assignment at level  $k-1$ . If  $i_{k-1}$  is assigned to  $\ell_{(i_{k-1})}$ , then  $i_{k-1}$  is removed from  $I$ ,  $\ell_{(i_{k-1})}$  is removed from  $J$ ,  $b_{i_{k-1} \ell_{(i_{k-1})}} = +\infty$ . Of course,

with this restriction, the assignment  $x_{i_k \ell(i_k)}$  is now possible, and so  $b_{i_k \ell(i_k)}$  is calculated in the usual manner. The process is then repeated. If on the other hand, the bound  $B'$  resulting from prohibiting the assignment  $i_k$  to  $\ell(i_k)$ , is less than  $C^*$ , which will be referred to as weak fathoming, then object  $i_k$  is assigned to some other unassigned location. This is discussed in more detail in the forward move of the search. The cases of strong and weak fathoming are depicted below.



#### Strong Fathoming

\* Continue at level k-1 by trying to assign object  $i_{k-1}$  to some unassigned location  $\neq \ell(i_{k-1})$ .



#### Weak Fathoming

\* Continue at level k by trying to assign object  $i_k$  to some unassigned location  $\neq \ell(i_k)$ ,

## B. Progress (Forward Move)

If the bound  $B$  on all completions of the partial solution at hand is less than  $C^*$ , then it is not impossible to find a completion of the partial solution with an objective less than  $C^*$ . At any rate, suppose that the current level of the tree is  $k$ . In other words, suppose that  $k$  objects have already been assigned, and we are trying to assign another object. Now we have a choice on which object to try to assign. Various possibilities exist. For example, we may choose  $i_{k+1}$  to be an unassigned object with maximum interactions with already assigned objects, or choose  $i_{k+1}$  to be an unassigned object with maximum interactions with most recently assigned object  $i_k$ . At any rate, we pick a suitable unassigned object  $i_{k+1}$ . This object is assigned to an unassigned location which is not prohibited for object  $i_{k+1}$ . This location  $\ell(i_{k+1})$  can be picked in such a way that the total weighted interaction with assigned objects is minimal, e.g. choose  $\ell(i_{k+1})$  which minimizes 
$$\sum_{j=1}^k u_{i_{k+1}i_j} d_{t\ell(i_j)} + \sum_{j=1}^k u_{i_j i_{k+1}} d_{\ell(i_j)t}$$
 over  $t/J$  such that  $x_{i_{k+1}t} = 1$  is not prohibited.

Choosing an object  $i_{k+1}$  and a location  $\ell(i_{k+1})$  in the above fashion, increases the level of the tree by 1. A bound on completing the partial solution is calculated as before and the process is repeated.

Needless to say, when the level of the tree is  $m$ , and if the cost is less than  $C^*$ , then  $C^*$  is replaced by the cost, and the corresponding assignment is stored.

## Termination

We have described forward and backward progress of the tree search. If the level of the tree ever reaches value zero, then we stop. This would

mean that we are currently at level one, and are trying to backtrack.

This means that all possible assignments under  $x_{i_1 \ell(i_1)} = 1$  and  $x_{i_1 \ell(i_1)} = 0$  have already been enumerated, i.e. all possible ways of assigning the  $m$  objects are enumerated, and we stop. The stored assignment and corresponding  $C^*$  give the optimal solution.

### Summary of the Algorithm

We have developed above all the details required to describe a specific solution procedure of the quadratic assignment problem.

#### Initialization Step

Let  $P(i) = \emptyset$  for  $i = 1, 2, \dots, m$  (prohibited locations). Let  $C^* = \infty$ .

Choose an object  $i_1$  and place it in location  $\ell(i_1)$ .  $i_1$  can be determined by maximizing  $\sum_{j=1}^m u_{ij}$  for  $i = 1, 2, \dots, m$ , and  $\ell(i_1)$  can be determined by minimizing  $\sum_{i=1}^n d_{ij}$  for  $j = 1, 2, \dots, n$ . Let  $I = \{i_1\}$  and  $J = \{j_1\}$ . Let  $k = 1$  and go to step 1.

#### Step 1 (Forward Move)

Calculate a lower bound  $B$  on all completions of current partial solution.

$B = C_1 + C_2 + C_3$ , where  $C_1$ ,  $C_2$ , and  $C_3$  are calculated as shown earlier, with the exception that  $b_{ij} = \infty$  if  $j \in P(i)$ . If  $B \geq C^*$  go to step 2.

Otherwise pick  $i_{k+1} \notin I$  such that  $u_{i_{k+1} i_k} = \max_{i \notin I} u_{ii_k}$ , and place  $i_{k+1}$  in

$\ell(i_{k+1})$  where  $\ell(i_{k+1})$  is determined by

$$\text{Minimizing } f_{i_{k+1} j} + \sum_{t=1}^k u_{i_{k+1} i_t} d_{j \ell(i_t)} + \sum_{t=1}^k u_{i_t i_{k+1}} d_{\ell(i_t) j} \\ \text{for } j \notin J, j \notin P(i_{k+1})$$

Replace  $I$  by  $I \cup \{i_{k+1}\}$  and  $J$  by  $J \cup \{\ell(i_{k+1})\}$ .

If  $k = n-1$ , calculate the new value of  $C_1$  (cost of complete assignment).

If  $C_1 < C^*$ , replace  $C^*$  by  $C_1$ , store  $(i_t, \ell(i_t))$  for  $t = 1, 2, \dots, n$ , replace

$k$  by  $n$  and go to step 2. If  $C_1 \geq C^*$  then replace  $k$  by  $n$  and go to Step

2. If  $k < n-1$ , then replace  $k$  by  $k+1$  and repeat Step 1.

## Step 2 (Fathom)

$i_k$  is removed from  $I$  and  $\ell(i_k)$  is removed from  $J$ . Replace  $P(i_k)$  by  $P(i_k) \cup \{\ell(i_k)\}$ . Calculate a lower bound  $B$  on completions of the partial solution  $x_{i_t \ell(i_t)} = 1$  for  $t = 1, 2, \dots, k-1$  and  $x_{i_k \ell(i_k)} = 0$ , where  $B = C_1 + C_2 + C_3$ .  $C_1, C_2$ , and  $C_3$  are calculated as mentioned above, with the exception that  $b_{ij} = +\infty$  if  $j \in P(i)$ . If  $B \geq C^*$  go to step 3. Otherwise assign  $i_k$  to a location  $t \notin J \cup P(i_k)$  such that the cost  $f_{i_k t} + \sum_{j=1}^{k-1} u_{i_k i_j} d_{t \ell(i_j)} + \sum_{j=1}^{k-1} u_{i_j i_k} d_{\ell(i_j) t}$  is minimized over  $t \notin J \cup P(i_k)$ .  $i_k$  is added to  $I$ , and  $t$ , the new  $\ell(i_k)$ , is added to  $J$ . Go to step 1.

## Step 3 (Strong Fathom)

$i_{k-1}$  is deleted from  $I$  and  $\ell(i_{k-1})$  is deleted from  $J$ .  $\ell(i_{k-1})$  is placed in  $P(i_{k-1})$  and  $P(i_k)$  is replaced by the empty set.  $k$  is replaced by  $k-1$ . If  $k = 0$  go to step 4, otherwise go to step 2.

## Step 4 (Termination)

Search has been completed. Optimal cost is  $C^*$  and its corresponding assignment  $(i_k, \ell(i_k))$ ,  $k = 1, 2, \dots, n$  is the optimal assignment. Stop.

The following points are helpful to keep in mind.

1. During the initialization step an upper bound  $C^* = \infty$  is used. As the search progresses,  $C^*$  denotes the objective of the best available complete assignment. Note that  $P(i)$  represents the locations that object  $i$  cannot be assigned to. These are initialized by the empty sets.
2. Step 1 represents a forward step, where the level of the tree increases by 1 unit. In this case the bound is less than  $C^*$ , and hence a complete solution with an objective better than  $C^*$  is possible.
3. Step 2 is a fathoming step, where  $B \geq C^*$ . In this case the last assignment is prohibited. Immediately a new bound is calculated. If the new bound is less than  $C^*$ , then a forward move is made. But if the bound is still greater than or equal to  $C^*$ , then a strong fathoming (Step 3) is made, and the level of the tree is reduced. Of course,

strong fathoming is most desirable, since it avoids the expensive task of trying to locate object  $i_k$  to a free location other than  $\ell(i_k)$ .

### Sub-Optimal Solutions

Due to its highly combinatorial nature, the task of finding and verifying the optimal solution within a reasonable computation time, for problems with  $n > 15$ , is almost impossible. In order to solve larger problems, we must resort to suboptimal solutions. Rather than using some heuristics, which may work very well for some problems, but which also may not be satisfactory for other problems, we will use the branch and bound procedure, itself, to obtain suboptimal solutions (or optimal) within a desired degree of accuracy from the optimal. We propose two methods for implementing this general strategy.

#### Method 1

Recall that a partial solution is fathomed, if the lower bound on all completions is at least as big as  $C^*$ , the best known objective. Suppose that a partial solution is fathomed if  $B \geq \alpha C^*$ , where  $\alpha \in (0,1]$ . In this case, the partial solution is abandoned if there is no hope that it leads to an objective which is better than  $\alpha C^*$ . To illustrate, suppose that  $C^*$  is 1000, and we choose  $\alpha = 0.95$ . Then we fathom if  $B \geq 950$ . This means that if the bound on completions of the partial solution cannot lead to an improvement of 50 units over what we already have, then it does not pay off to pursue the partial solution. The objective of this simple strategy is clear. We want to fathom, i.e. abandon partial solutions, quickly, even if they lead to a slight improvement. As the computational experience shows, this simple strategy speeds the computational effort considerably. Of course, as a new  $C^*$  is found, then we fathom whenever the bound is greater or equal to  $\alpha$  times the new  $C^*$ . The pro-



cedure continues until we cannot find a feasible solution with an objective less than  $\alpha C^*$ . So we have a feasible assignment with objective  $C^*$ , coupled with the statement that the optimal objective is greater than or equal to  $\alpha C^*$ .

#### Choice of $\alpha$

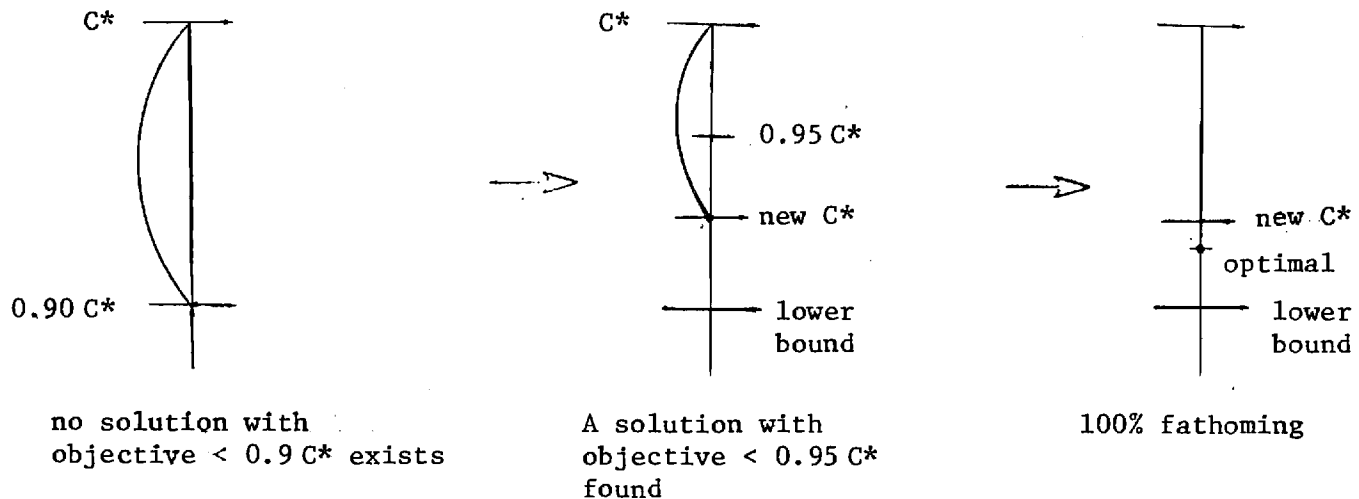
Of course, if  $\alpha$  is small, then fathoming will speed up considerably, resulting in a small computational effort. But on the other hand, the quality of the best obtained feasible solution is not guaranteed. To illustrate, suppose that at termination  $C^*$  is 1000 and suppose that  $\alpha = 0.60$ . All we can say in this case is that we have a feasible solution with objective 1000, and the true optimal has an objective greater or equal to 600. This, of course, is not satisfactory. Suppose, however, that we used  $\alpha = 0.95$ . Then if at termination  $C^*$  is 1000, we can say that the optimal is greater than or equal to 950. This statement is indeed much stronger than the previous statement.

From this discussion, it is clear that there is a tradeoff between the computational effort and the value of  $\alpha$ . We recommend values of  $\alpha \geq 0.9$  depending on the accuracy required.

#### Exact Solution By Stepped Fathoming

A slight modification of the above scheme can be used to find the exact optimal solution. Suppose that an initial  $\alpha_1$  is used, say  $\alpha_1 = 0.90$ . Eventually we achieve a final  $C^*$  and cannot find a feasible solution with objective less than  $0.90 C^*$  i.e.  $0.9 C^*$  is a lower bound on all solutions. We either stop here, or else switch to  $\alpha_2 \geq \alpha_1$ , say  $\alpha_2 = 0.95$ . We start the search from the stored best solution corresponding to  $C^*$ , which also includes all the prohibited locations. We are either able to find a feasible solution with objective less than  $0.95 C^*$  (none less than  $0.9 C^*$  exist), and repeat the process, or else conclude that none exists. We may then

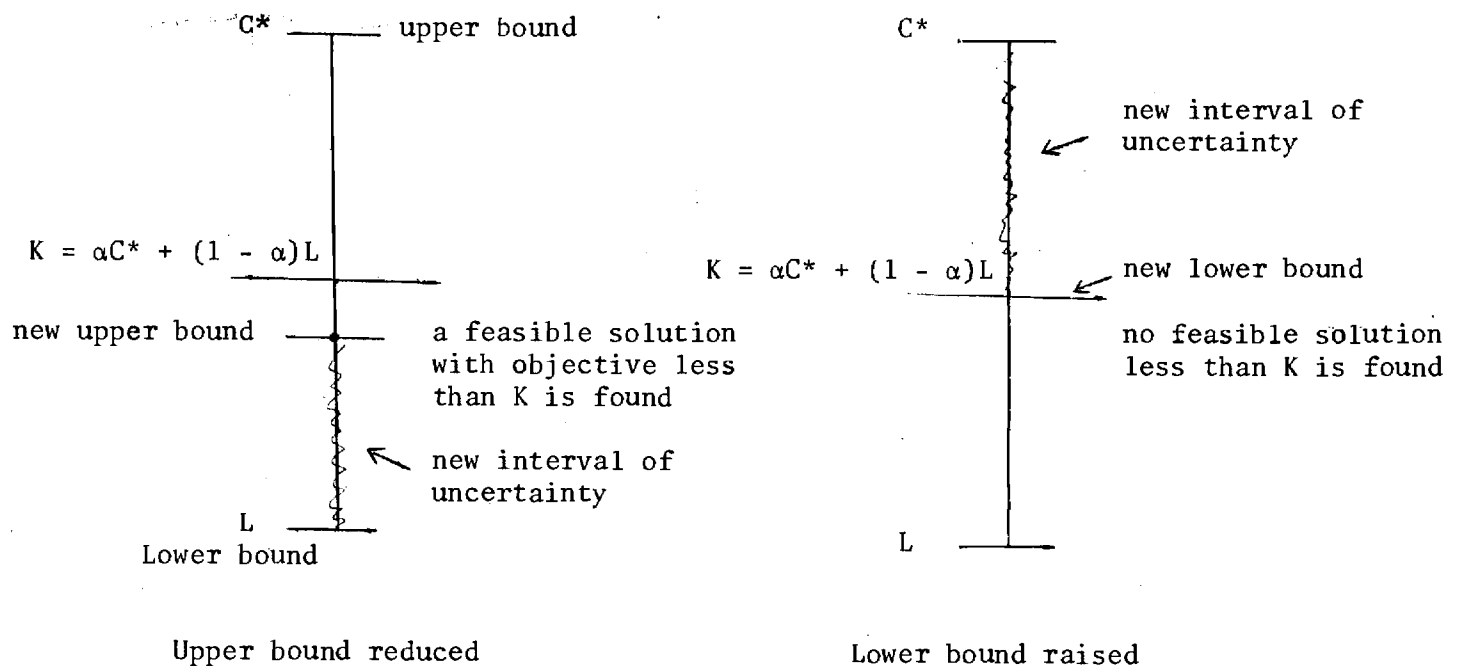
switch to 0.97 fathoming and eventually 100% fathoming. This of course, will lead to the optimal solution. Even though parts of the tree will be searched more than once, the quick fathoming outweighs some repeated efforts.



## Method 2

This method again fathoms whenever the bound  $B$  is greater than or equal to  $K$ , where  $K$  is related to, but not necessarily equal to  $C^*$  ( $K \leq C^*$ ). Again, we have the same interpretation as before. A partial solution is pursued if it guarantees a nontrivial improvement over what we have got. We will describe a procedure to find a suitable  $K$ . At each stage of the algorithm, we have an upper bound (optimal  $\leq$  upper bound), namely  $C^*$ . A lower bound on the overall problem  $L$ , can be devised. At each stage of the algorithm, we have a feasible solution with objective  $C^*$ , and we know that the optimal must be greater than or equal to  $C^*$ . Rather than fathoming on  $C^*$ , suppose we fathom on  $K = \alpha C^* + (1 - \alpha)L$ , where  $\alpha \in [0, 1]$ . Since  $L < C^*$ , then  $K = \alpha C^* + (1 - \alpha)L \leq C^*$ . Two cases are possible. In the first case, we will be able to find a complete solution with objective less than  $K$ . The objective of this new solution becomes the new upper bound  $C^*$  and the process is repeated. In the second case, we will not be able to find such a solution. This automatically implies

that there are no solutions with objective less than  $K$ , and hence  $K$  itself is the new lower bound. The process is repeated. From this we keep narrowing the gap between the lower and upper bounds, either by lowering the upper bound (finding an improved feasible assignment) or by raising the lower bound (by finding that there is no feasible solution with objective lower than the current lower bound). When the difference between the lower bound, and the current best objective value (upper bound) is smaller than a prescribed tolerance, we either stop, or switch the fathoming to  $K = C^*$ . Of course, if the tolerance is zero, or if we switch the fathoming to  $C^*$  (i.e., switch  $\alpha$  to 1), then we will find the true optimal solution.



#### Choice of $\alpha$

$\alpha$  is any number in the interval  $[0,1]$ . Of course, if  $\alpha$  is close to 1, then we are in effect fathoming on a number very close to  $C^*$ , and the search will not speed considerably. On the other hand, if  $\alpha$  is close to zero, then improvement is achieved only if we obtain a feasible solution

very close to the overall lower bound. In this case, fathoming will be fast, but it is likely not to obtain feasible solutions less than  $K$ . So again a tradeoff is required here. We recommend values of  $\alpha$  equal to 0.5 so that the interval of uncertainty is halved at each time.

#### Calculation of an initial overall lower and upper bounds

An initial lower bound, and also a reasonable (not equal to  $\infty$ ) initial upper bound are needed for the above procedure. To calculate the lower bound, first calculate a lower bound  $b_{ij}$  on the cost of locating object  $i$  to location  $j$ , for  $i = 1, 2, \dots, m$  and  $j = 1, 2, \dots, n$ . The interactions between object  $i$  and all other objects are put in a descending order, and similarly, the distances from location  $j$  and all other locations, are put in an ascending order. The inner product of the two vectors give  $b_{ij}$ . Then a linear assignment problem is solved to find  $L$ . More precisely, let  $L$  be the optimal objective of the following problem.

$$\begin{aligned} \text{Minimize} \quad & \sum_{j=1}^n \sum_{i=1}^m b_{ij} x_{ij} \\ & \sum_{j=1}^n x_{ij} = 1 \quad i = 1, 2, \dots, m \\ & \sum_{i=1}^m x_{ij} \leq 1 \quad j = 1, 2, \dots, n \\ & x_{ij} \geq 0 \quad \text{all } i \text{ and } j \end{aligned}$$

An upper bound  $C^*$  is immediately available, by calculating the quadratic cost of the optimal assignment resulting from the above problem. Now Method 2 can be initiated.

#### Preassigned Objects

In many applications, a large number of objects is already preassigned, and only some new objects are to be placed, such that the overall cost

is minimized (e.g. some plants are already assigned, and it is desirable to assign some new plants). In this case, the algorithm can be applied with few obvious modifications in the calculations. Since the preassigned objects and their locations will always be assigned, then these objects will always be in the set  $I$  and their locations will always be in the set  $J$ . In the tree search, if  $\gamma$  objects are already assigned, we start the search by assigning more objects, i.e. the level of the tree starts at  $\gamma+1$ . If the level of the tree ever becomes  $\gamma$ , then we stop. If an overall lower bound on all feasible solutions to the problem is desired, we recommend calculating  $C_1 + C_2 + C_3$  as before, where  $I$  and  $J$  are the indices of the preassigned objects and their locations.

#### A Heuristic For Fixing Locations Of Some Key Objects

For large problems, obtaining an optimal solution, or even a sub-optimal solution, with a good accuracy, may be prohibited due to the fact that the computational effort grows very fast as a function of the problem size. One of the difficulties of the quadratic assignment problem is that many solutions exist which are either optimal or very close to the optimal and hence, a large portion of the decision tree must be explicitly enumerated if the true optimal is to be found. This disadvantage can be significantly reduced by preassigning some objects in a seemingly good set of locations. These objects are then fixed once and for all, and the search continues on the remaining objects. If we choose the objects which are preassigned in a logical manner, and so their preassigned locations, even though we may miss the true optimal, it is likely that we will be very close to it, provided that the other objects are assigned optimally, once the preassigned objects are fixed. In a sense, the branch and bound procedure will tell us how to optimally assign the remaining objects, once we commit ourselves to preassigning some objects. If the

preassignment was done intelligently, we will likely be very close to the optimal solution, and meanwhile accomplish the following two tasks:

1. The tree level is reduced, thus drastically reducing the computational effort.
2. By fixing the location of some key objects, the lower bounds calculated will be tighter, which will allow faster fathoming, again resulting in reduced computational effort.

#### Basis for Choosing Preassigned Objects

Objects which interact with a large number of objects, and which have large total interactions, are desired to be assigned in "central" locations, so that they would be close to many possible objects. This remark can be used effectively to fix the locations of some objects. We can rank the objects according to a combination of;

1. The number of objects to which any given fixed object is interacting
2. The total number of interactions from any given object and all other objects

The locations are ranked according to their centralization, e.g. location  $i$  is ranked according to  $\sum_{j=1}^n d_{ij}$ . The best location  $i$  is the one with smallest  $\sum_{j=1}^n d_{ij}$ , and so forth. A possible way of fixing key objects is to assign the best ranking object to the best ranking location. Then the object which has the most number of interactions to the most recently assigned object is placed in a location which minimizes total cost. Then the next best ranking object can be assigned to a location which minimizes total cost, and so forth. This process is continued until a certain percentage of the objects are assigned, say 20%. We then seek improving the quality of the preassignment by rotating the positions of these objects. Then we go to the branch and bound scheme. This procedure has proved effective in obtaining good quality solutions.

### Calculation Of Overall Lower Bounds In The Preassigned Case

We have discussed earlier the importance of fixing some objects to certain locations apriori. The branch and bound algorithm will then determine the optimal completion of the partial apriori fixed solution. We argued earlier, that if these fixed objects and their locations, are chosen in a logical manner, it is likely that the optimal not be missed by a large amount. It may be helpful, however, to calculate an overall lower bound, on the portion of the decision tree, which we did not enumerate, due to the initial partial solution.

To illustrate, suppose that object  $i$  is placed in location  $\ell(i)$ , once and for all during the computation, for  $i = 1, 2, \dots, k$ . We would like to calculate a lower bound on the portion of the tree where not all the objects  $1, 2, \dots, k$  are assigned to  $\ell(1), \ell(2), \dots, \ell(k)$ .  $2^k - 1$  cases must be investigated, since any given object  $i$ , can either be placed in  $\ell(i)$  or at some other location. For each such case, a lower bound is calculated. An overall lower bound, on the unenumerated portion of the tree is then the minimal of these lower bounds. The lower bound, where object  $i$  is placed in  $\ell(i)$  for  $i \in I \subset \{1, 2, \dots, k\}$ , and object  $i$  is prohibited from location  $\ell(i)$  for  $i \in \{1, 2, \dots, k\} - I$ , can be calculated as follows.

$$LB = C_1 + C_2 + C_3$$

where

$C_1$  is the cost of interaction among objects in  $I$ .

$$C_1 = \sum_{i,j \in I} u_{ij} d_{\ell(i), \ell(j)} + \sum_{i \in I} f_i \ell(i)$$

$C_2$  is the cost of interaction from unassigned objects to objects in  $I$

$C_3$  is the cost of interaction from unassigned objects to unassigned objects.

$C_2$  and  $C_3$  can be calculated by solving the following linear assignment problem.

$$\begin{aligned}
& \text{Minimize} && \sum_{\substack{j=1 \\ j \notin J}}^n \sum_{\substack{i=1 \\ i \in I}}^m b_{ij} x_{ij} \\
& \text{Subject to:} && \sum_{\substack{j=1 \\ j \notin J}}^n x_{ij} = 1 && \text{for } i \in \{1, 2, \dots, m\} - I \\
& && \sum_{\substack{i=1 \\ i \in I}}^m x_{ij} \leq 1 && \text{for } j \in \{1, 2, \dots, n\} - J \\
& && x_{ij} \geq 0 && \begin{aligned} & i \in \{1, 2, \dots, m\} - I \\ & j \in \{1, 2, \dots, n\} - J \end{aligned}
\end{aligned}$$

where  $J = \{j : j = \ell(i) \text{ for some } i \in I\}$ , and  $b_{ij}$  is a bound on the cost of assigning object  $i$  to location  $j$ .  $b_{ij}$  is calculated as follows.

$$\begin{aligned}
b_{i\ell(i)} &= \infty && \text{if } i \in \{1, 2, \dots, k\} - I \\
b_{ij} &= c_2 + c_3 && \text{otherwise}
\end{aligned}$$

where

$$\begin{aligned}
c_2 &= \sum_{t \in I} u_{it}^d \ell(t)_j + \sum_{t \in I} u_{ti}^d \ell(t)_j + f_{ij} \\
c_3 &= v \cdot w
\end{aligned}$$

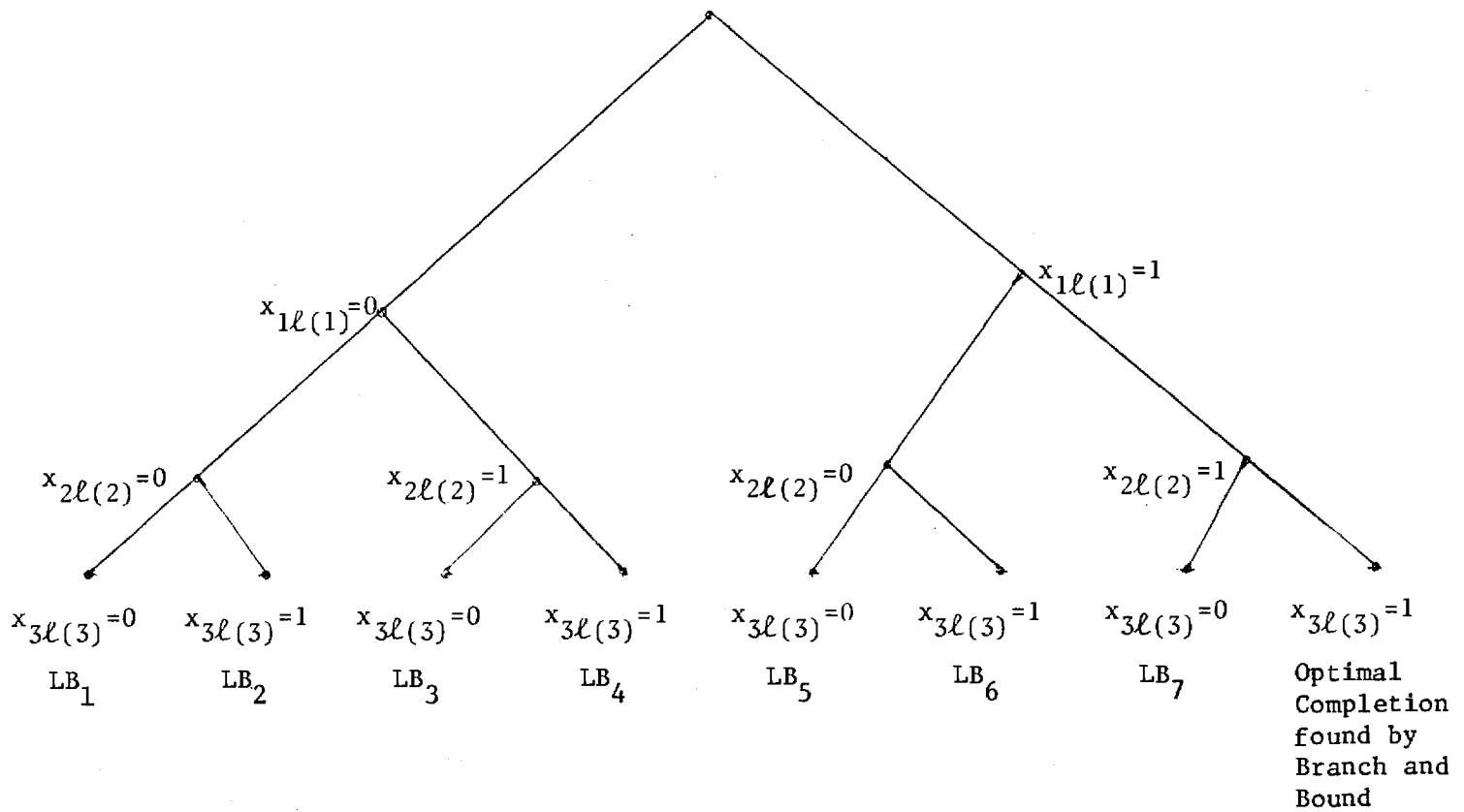
$v$  is a vector of interactions from object  $i$  to objects  $\{1, 2, \dots, m\} - I$ , and from objects  $\{1, 2, \dots, m\} - I$  to object  $i$ , arranged in descending order.

$w$  is a vector of distances from location  $j$  to locations  $\{1, 2, \dots, n\} - J$ , and from locations  $\{1, 2, \dots, n\} - J$  to location  $j$ , arranged in an ascending order. If  $n \neq m$ , complete the vector  $v$  by adding zeros from the bottom so that  $v$  and  $w$  have the same size.

After solving the above linear assignment problem, we get a bound on all solutions where the assignments  $i$  in  $\ell(i)$  for  $i \in I \subset \{1, 2, \dots, k\}$  is enforced, and the assignments  $i$  in  $\ell(i)$  for  $i \in \{1, 2, \dots, k\} - I$  are prohibited. Lower bounds on all possible ways of choosing  $I \subset \{1, 2, \dots, k\}$  are calculated, and the minimal represents the absolute lower bound on the overall problem. The case of  $k = 3$  is illustrated below. The optimal completion of  $x_{i\ell(i)} = 1$  for  $i = 1, 2, 3$  is



found by the algorithm.  $LB_1, \dots, LB_7$  are calculated as discussed above,  
 and the overall lower bound on the unenumerated portion of the tree is  
 then  $\min_{1 \leq j \leq 7} LB_j$ .



### Section III

#### THE TRAVELING SALESMAN PROBLEM

In this section we will present a branch and bound algorithm for solving nonsymmetric as well as symmetric traveling salesman problems.

As a first step, we formulate and solve the dual of the traveling salesman problem. If the optimal dual solution results in an oriented tour (or just a tour in the symmetric case), then we also have an optimal solution of the traveling salesman problem. Otherwise, we at least obtain a right lower bound on the optimal solution of the traveling salesman problem, and a new distance matrix which is "well-conditioned".

A branch and bound procedure, similar to that of the general quadratic assignment model, is devised to find the optimal tour, i.e., to close any gap between the primal and dual problems. The major simplification obtained for the special case of the traveling salesman problem is that we start the branch and bound search from an infeasible solution which is close to the optimal tour, thus considerably reducing the computational effort. Of course, this is made possible by the use of duality in obtaining a tight lower bound. The procedure, which is suitable for both symmetric and nonsymmetric traveling salesman problems, is an extension of the method of Held and Karp for solving the symmetric problem.

#### Formulation of the Problem and its Dual

Suppose that a salesman is given a list of  $n$  cities, where each pair of cities,  $i$  and  $j$ , is connected by a link of length  $c_{ij}$  (if cities  $i$  and  $j$  are not directly linked, let  $c_{ij} = +\infty$ ). The problem is then to find

a tour of minimal length which the salesman can follow so as to visit each city exactly once, and then return to where he started.

Let  $x_{ij}$  be a variable relating to link  $ij$ .  $x_{ij}$  assumes a value of 1 or a value of 0. If the salesman travels from city  $i$  to city  $j$ , then  $x_{ij} = 1$ , otherwise  $x_{ij} = 0$ . The problem can then be stated as follows:

$$\text{Minimize} \quad \sum_{\substack{j=1 \\ j \neq i}}^n \sum_{i=1}^n c_{ij} x_{ij}$$

$$\text{Subject to:} \quad \sum_{\substack{j=1 \\ j \neq i}}^n x_{ij} = 1 \quad i = 1, 2, 3, \dots, n \quad (1)$$

$$\sum_{\substack{i=1 \\ i \neq j}}^n x_{ij} = 1 \quad j = 1, 2, 3, \dots, n \quad (2)$$

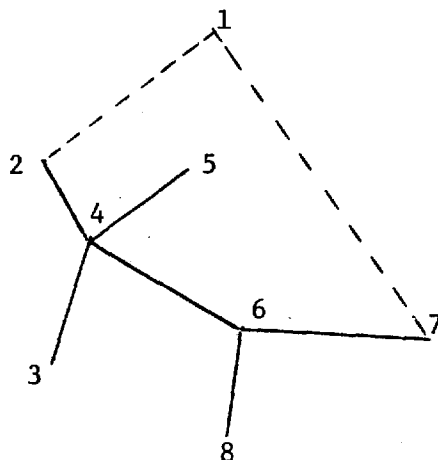
$$x_{ij} \text{ is 0 or 1} \quad i = 1, \dots, n; j = 1, \dots, n \quad (3)$$

No subtours

Since the restrictions (1), (2), and (3), by themselves, may allow the meaningless answer of having two or more disjoint subtours, a restriction eliminating subtours must be explicitly introduced. There are various methods in the literature for enforcing this restriction. We will adopt the one-tree method of Held and Karp, which if coupled with (1), (2), and (3), will ensure the elimination of subtours.

Let us begin by reviewing the definition of a 1-tree [10]. A 1-tree of the graph composed of the  $n$  cities and their interconnecting links is a tree which spans  $n-1$  cities, coupled with two distinct links from the remaining city, to any two of the  $n-1$  cities spanned by the tree. Therefore, by definition, a 1-tree contains exactly one subtour, or one

cycle. Of course, the number of links in the cycle is less than or equal to  $n$ . An example of a 1-tree is illustrated below. Notice that the 1-tree is composed of a tree spanning cities (nodes) 2 through 8, plus two links from node 1 to nodes 2 and 7.



An Example of a 1-Tree

Now recall that constraints (1), (2), and (3) either admits no subtours, or else at least two subtours. By definitions of the one-tree, it admits exactly one subtour (which may be a complete tour if it has exactly  $n$  links). Therefore constraints (1), (2), and (3), coupled with the constraint that the vector  $x$  forms a 1-tree, admit no subtours. Therefore, the traveling salesman problem can be formulated as follows:

$$\begin{aligned}
 &\text{Minimize} && \sum_{\substack{j=1 \\ j \neq i}}^n \sum_{i=1}^n c_{ij} x_{ij} \\
 &\text{Subject to:} && \sum_{\substack{j=1 \\ j \neq i}}^n x_{ij} = 1 && i = 1, 2, \dots, n \\
 &&& \sum_{\substack{i=1 \\ i \neq j}}^n x_{ij} = 1 && j = 1, 2, \dots, n \\
 &&& x \in X
 \end{aligned}$$

where  $X$  is the set of 1-trees spanning the graph, i.e.,

$$X = \{(x_{11}, \dots, x_{1n}, \dots, x_{n1}, \dots, x_{nn}) : x_{ij} = 0 \text{ or } x_{ij} = 1 (i \neq j), \\ \text{and the links } ij \text{ with } x_{ij} = 1 \text{ form a 1-tree}\}.$$

Note that there are very efficient procedures for finding a 1-tree with minimal cost [5, 13]. This will involve pulling one node out of the  $n$  nodes, which we will refer to as the initial node, and then:

1. finding a tree of minimal length which spans the remaining  $n-1$  nodes
2. finding two links of minimal length which connect the initial node to the spanning tree.

Any node can be used as the initial node, and we will discuss later a procedure for making a "good" choice of the initial node. The calculation of the 1-tree with minimal cost, in the presence of restrictions (1), (2), and (3), namely the traveling salesman problem, is indeed a difficult problem. If we handle the restrictions (1) and (2) via Lagrangian multipliers (dual variables) and put them into the objective function, we are left with the very simple problem of finding a minimal 1-tree. The results of the minimal 1-tree can be used to update the Lagrangian multipliers. This leads to the following dual formulation.

#### Dual Problem

Maximize  $\theta(u, v)$

$u$  and  $v$  are unrestricted

where,

$$\begin{aligned} \theta(u, v) &= \text{Minimum}_{x \in X} \left\{ \sum_j \sum_i c_{ij} x_{ij} + \sum_i u_i \left( \sum_j x_{ij} - 1 \right) + \sum_j v_j \left( \sum_i x_{ij} - 1 \right) \right\} \\ &= \text{Minimum}_{x \in X} \left\{ \sum_j \sum_i (c_{ij} + u_i + v_j) x_{ij} - \sum_i u_i - \sum_j v_j \right\} \end{aligned}$$

$$= -\sum_i u_i - \sum_j v_j + \text{Minimum}_{x \in X} \left\{ \sum_j \sum_i (c_{ij} + u_i + v_j) x_{ij} \right\}$$

Note that evaluating  $\theta(u,v)$ , for fixed values of  $u$  and  $v$  is an easy task since it involves finding a minimal l-tree, where the cost  $c_{ij}$  of the link  $ij$  is replaced by  $c_{ij} + u_i + v_j$ . Also note that in finding the minimal l-tree, we are not concerned about the orientation of the links. In other words we are actually searching for a minimal l-tree, rather than an aborescence. This simplifies the problem of evaluating  $\theta(u,v)$  considerably. Of course, we may think of the vectors  $u$  and  $v$  as penalties for violating the proper orientation of the links. So if when calculating the minimal l-tree, too many links "left" node  $i$ , we will attempt to "penalize" this deficiency in the next iteration by increasing  $u_i$ . Similarly, if too many links "entered" node  $j$ , then we will attempt to "penalize" this deficiency by increasing  $v_j$ .

#### Simplification of the Symmetric Case

In the symmetric case, as long as the vector  $x$  forms a l-tree, and since the orientation of links is not important, we no longer require that the number of links leaving node  $i$  is 1, and the number of links entering node  $i$  is 1, so long as the number of links incident to node  $i$  is equal to 2. The primal problem can thus be reformulated as follows:

$$\begin{aligned} \text{Minimize} \quad & \sum_{\substack{j=1 \\ j \neq i}}^n \sum_{i=1}^n c_{ij} x_{ij} \\ \text{Subject to:} \quad & \sum_{\substack{j=1 \\ j \neq i}}^n x_{ij} + \sum_{\substack{j=1 \\ j \neq i}}^n x_{ji} = z \quad \text{for } i = 1, 2, \dots, n \\ & x \in X \end{aligned}$$

where  $X$  is the collection of l-trees. The above restrictions ensure

forming a loop. If the loop is not "oriented", it can be made so, by simply reorienting some of the links, without affecting the cost. The dual of the above problem is given below, with only one set of dual variables given by the vector  $u$ .

Maximize  $\theta(u)$

$u$  is unrestricted

where,

$$\begin{aligned}\theta(u) &= \text{Minimum}_{x \in X} \sum_{\substack{j=1 \\ j \neq i}}^n \sum_{i=1}^n c_{ij} x_{ij} + \sum_{i=1}^n u_i \left( \sum_{\substack{j=1 \\ j \neq i}}^n x_{ij} + \sum_{\substack{j=1 \\ j \neq i}}^n x_{ji} - 2 \right) \\ &= -2 \sum_{i=1}^n u_i + \text{Minimum}_{x \in X} \sum_{\substack{j=1 \\ j \neq i}}^n \sum_{i=1}^n (c_{ij} + u_i + u_j) x_{ij}\end{aligned}$$

From now on we will concentrate on the general nonsymmetric case with dual variables  $u$  and  $v$ . For the symmetric case, the dual problem can be simplified as illustrated above, with no need for the dual variable  $v$ .

#### Interpretation of the Dual Problem

Let  $f(x) = cx$ ,  $g(x)$  and  $h(x)$  be  $n$ -vector functions defined as follows:

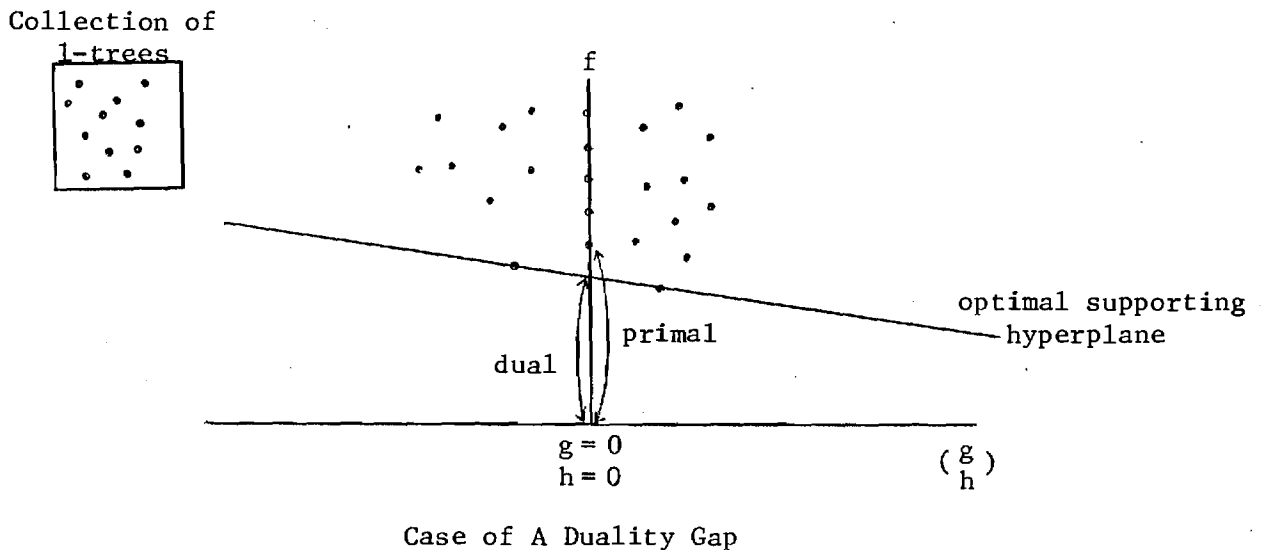
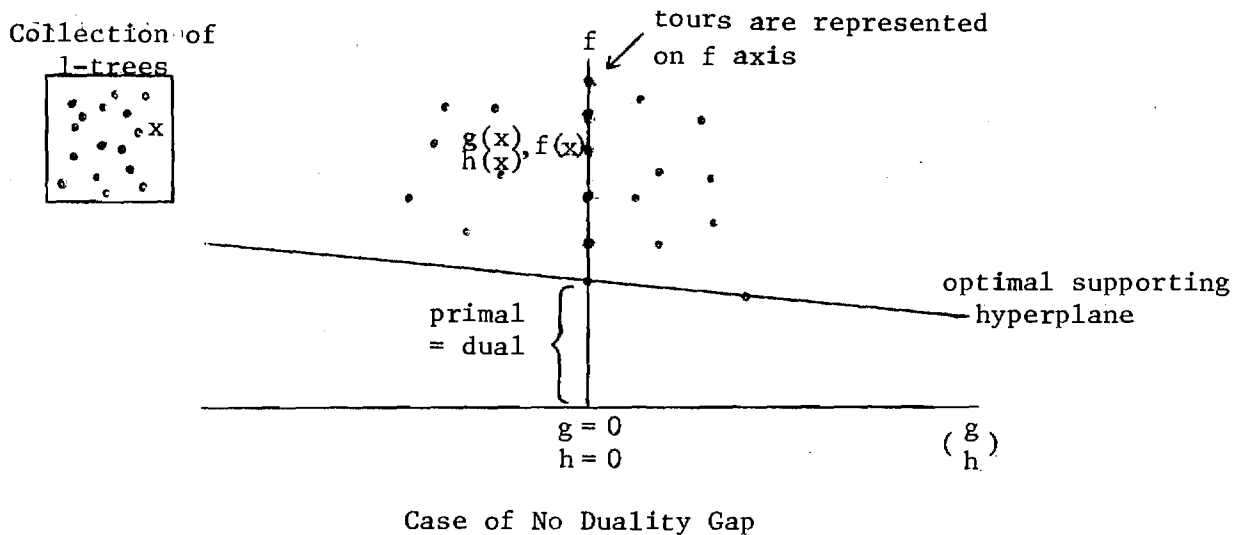
$$g_i(x) = \sum_{\substack{j=1 \\ j \neq i}}^n x_{ij} - 1 \quad \text{and} \quad h_i(x) = \sum_{\substack{j=1 \\ j \neq i}}^n x_{ji} - 1 \quad \text{for } i = 1, 2, \dots, n.$$

The dual problem can be interpreted on the  $\left(\begin{pmatrix} g \\ h \end{pmatrix}, f\right)$  plane. Each 1-tree in  $X$  can be mapped into the  $\left(\begin{pmatrix} g \\ h \end{pmatrix}, f\right)$  plane as  $\left(\begin{pmatrix} g(x) \\ h(x) \end{pmatrix}, cx\right)$ . Doing this for all 1-trees, we can construct the image of  $X$  in the  $\left(\begin{pmatrix} g \\ h \end{pmatrix}, f\right)$  plane, which will be composed of discrete points. Since a feasible tour is a 1-tree with  $g(x) = h(x) = 0$ , then all feasible tours lie on the  $f$  axis, and the minimal feasible tour is the one which is "lowest" on the  $f$  axis. Now given  $(u, v)$ ,  $\theta(u, v) = \text{Minimum}_{x \in X} cx + ug(x) + vh(x)$ . Therefore, evaluating  $\theta$  corresponds to finding a hyperplane with slope  $\begin{pmatrix} -u \\ -v \end{pmatrix}$

which supports the set of discrete points in the  $\left(\left(\begin{smallmatrix} g \\ h \end{smallmatrix}\right), f\right)$  plane, and which has minimal intercept on the  $f$  axis. Therefore, the dual problem can be stated as follows:

Find the hyperplane which supports the image of  $X$  in the  $\left(\left(\begin{smallmatrix} g \\ h \end{smallmatrix}\right), f\right)$  plane whose minimal intercept on the  $f$  axis is maximal.

Two cases are possible. In the first case, the optimal hyperplane supports the set at a point on the  $f$  axis, i.e., at the optimal primal tour, in which case there is no duality gap. In the second case, the optimal supporting hyperplane does not pass through the optimal tour, in which case a gap exists.





### Some Properties of the Dual Function $\theta$

We will now study the dual function  $\theta$ . It turns out that  $\theta$  is concave and piecewise linear in  $u$  and  $v$ . This will help us determine how to modify the dual variables in such a way to maximize  $\theta$ .

Let the collection of 1-trees, with a fixed initial node, be  $x_1, x_2, \dots, x_t$ .

Then  $\theta(u,v)$  can be written as follows:

$$\begin{aligned}\theta(u,v) &= \text{Minimum}_{x \in X} \sum_i \sum_j c_{ij} x_{ij} + \sum_i u_i \left( \sum_j x_{ij} - 1 \right) + \sum_j v_j \left( \sum_i x_{ij} - 1 \right) \\ &= \text{Minimum}_{x \in X} cx + ug(x) + vh(x) \\ &= \text{Minimum}_{1 \leq j \leq t} cx_j + ug(x_j) + vh(x_j)\end{aligned}$$

It is clear that  $\theta$  is the minimum of a finite number of affine functions in  $u$  and  $v$ , namely  $cx_j + ug(x_j) + vh(x_j)$  for  $j = 1, 2, \dots, t$ . The dual function is illustrated below, and its concavity and piecewise linearity is clear from its construction. Note that each of the hyperplanes defining  $\theta$  corresponds to a 1-tree.

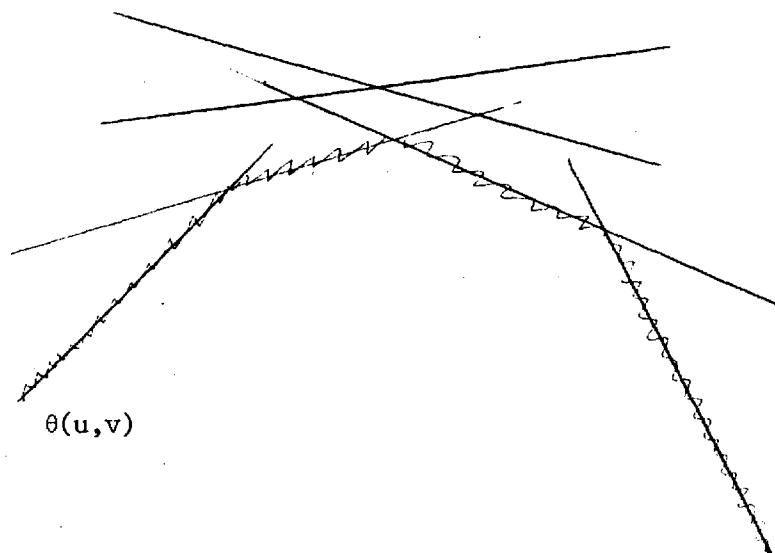


Illustration of the Dual Function

If we examine the  $(u,v)$  space, then it will be divided in regions

$R_1, R_2, \dots, R_t$ , where in region  $R_j$ ,  $\theta(u,v) = cx_j + ug(x_j) + vh(x_j)$ .

Note that some regions may be empty, which means that the corresponding 1-tree is never minimal regardless of  $u$  and  $v$ . At any rate, each  $R_j$  is in a polyhedral set, and for any  $(u,v)$  in the intersection of  $R_i$  and  $R_j$ , we have,

$$\theta(u,v) = cx_i + ug(x_i) + vh(x_i) = cx_j + ug(x_j) + vh(x_j)$$

Also, note that in the interior of region  $R_j$ ,  $\theta$  is differentiable

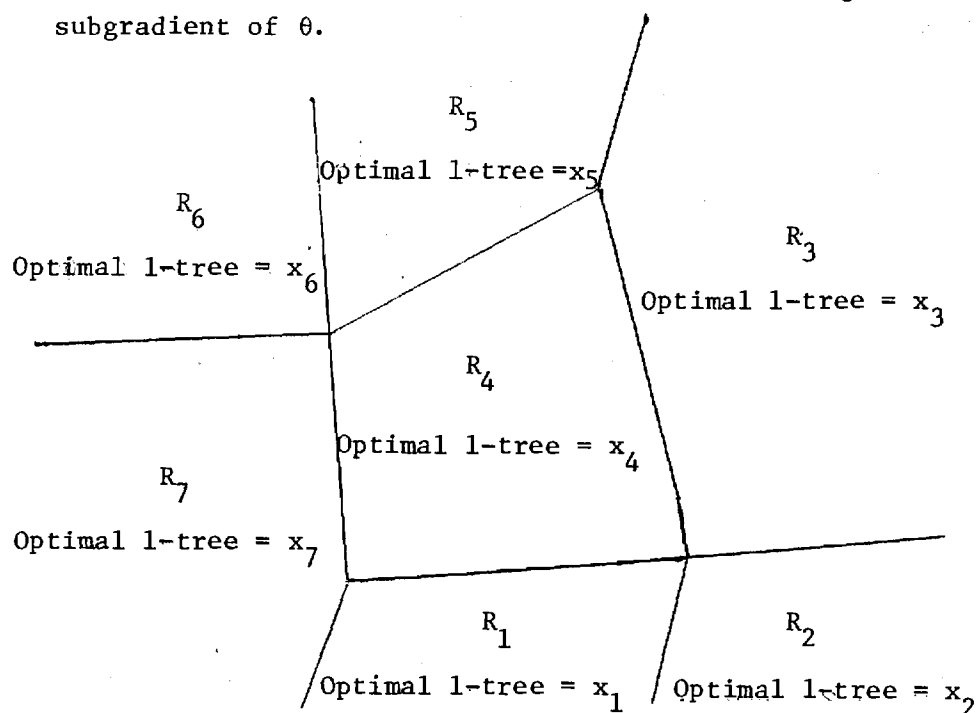
(actually linear) and has gradient  $\begin{pmatrix} g(x_j) \\ h(x_j) \end{pmatrix}$ .  $\theta$  is not differentiable

at the intersection of more than one region, but is subdifferentiable there, and the subgradients there can be characterized as con-

vex combinations of the gradients in these regions. In particular,

if  $(u,v) \in \bigcap_{j \in I} R_j$ , then  $\theta$  is subdifferentiable at  $(u,v)$  and

any subgradient is characterized by  $\sum_{j \in I} \lambda_j \begin{pmatrix} g(x_j) \\ h(x_j) \end{pmatrix}$  where  $\sum_{j \in I} \lambda_j = 1$  and  $\lambda_j \geq 0$  for each  $j \in I$ . In particular, each  $\begin{pmatrix} g(x_j) \\ h(x_j) \end{pmatrix}$  for  $j \in I$  is a subgradient of  $\theta$ .



Regions in the  $(u,v)$  Space

### Some Strategies for Maximizing $\theta$

There are various strategies for maximizing the dual function  $\theta$ , which makes good use of its special structure. We will discuss some alternative possibilities. Of course, the following procedures optimize the dual, but not necessarily the primal. In the absence of a duality gap, the optimal tour will be obtained in the process. Otherwise, as we will discuss later, a branch and bound scheme will be used. We summarize below three strategies for maximizing the dual function  $\theta$ . This is then followed by a more detailed discussion of each procedure.

#### Strategy 1    Tangential Approximation (or Decomposition)

The dual function  $\theta$  is approximated by only a portion of the affine functions defining it, and at each iteration the optimal  $(u,v)$  of the "tangential approximation" is found. Finding the shortest 1-tree either ends in verifying optimality or in adding another hyperplane, resulting in a better approximation of  $\theta$ .

#### Strategy 2    Steepest Ascent Optimization

At each point  $(u,v)$ , the steepest ascent direction of  $\theta$  is first found, then we optimize along this direction.

#### Strategy 3    Subgradient Optimization

The dual function is maximized by moving along subgradient vectors. In most cases this results in improving  $\theta$ , and/or getting closer to the optimal point. At each iteration a new subgradient is found by finding the shortest 1-tree.

### Strategy 1    Tangential Approximation (or Decomposition)

The decomposition principle can be used for maximizing the dual function. It may be helpful to rewrite the dual problem in a linear programming format, where  $z$  replaces  $\theta(u,v)$ .

#### Dual Problem

Maximize         $z$

Subject to:     $z \leq cx_j + ug(x_j) + vg(x_j)$         for  $j = 1, 2, \dots, t$

$z, u, v$  unrestricted in sign

The above dual problem is equivalent to its own dual. Denoting the dual variables by  $\lambda_j$ , the above problem can be alternatively represented as follows:

Minimize         $\sum_{j=1}^t \lambda_j (cx_j)$

Subject to:     $\sum_{j=1}^t \lambda_j g(x_j) = 0$

$\sum_{j=1}^t \lambda_j h(x_j) = 0$

$\sum_{j=1}^t \lambda_j = 1$

$\lambda_j \geq 0$          $j = 1, 2, \dots, t$

Since the number of 1-trees is very large, the dual problem has too many rows, and; needless to say, its own dual, has too many columns. Hence, the decomposition principle can itself be used as a row generation, or column generation technique for solving the dual problem. In particular, suppose that only  $k$  1-trees, say  $x_1, \dots, x_k$  are available. Then we can solve a master problem, which generates new dual variables  $(u_k, v_k)$ .

These in turn are passed to a subproblem which finds a new 1-tree.

An optimality test can be performed to test whether the optimal solution to the dual problem has been obtained. The following steps are typical of such an algorithm.

### Initialization Step

Find an initial feasible tour; a suitable heuristic can be used to find such a solution. This tour is  $x_1$ . Let  $k = 1$  and go to Step 1.

### Step 1 (Master Problem)

Solve the following problem.

$$\text{Minimize} \quad \sum_{j=1}^k \lambda_j (cx_j)$$

$$\text{Subject to:} \quad \sum_{j=1}^k \lambda_j g(x_j) = 0$$

$$\sum_{j=1}^k \lambda_j h(x_j) = 0$$

$$\sum_{j=1}^k \lambda_j = 1$$

$$\lambda_j \geq 0 \quad \text{for } j = 1, 2, \dots, k$$

Let  $z_k = \sum_{j=1}^k \lambda_j (cx_j)$ , and let  $u_k$  and  $v_k$  be the dual variables of the first two sets of constraints, respectively (the dual of the above problem can be solved, but the above problem is used for computational convenience).

### Step 2 (Subproblem)

Find the 1-tree of minimal length by solving the following subproblem.

Minimize  $(cx + u_k g(x) + v_k h(x))$   
 $x \in X$

Let the optimal be  $x_{k+1}$ . Go to Step 3.

### Step 3 (Optimality Test)

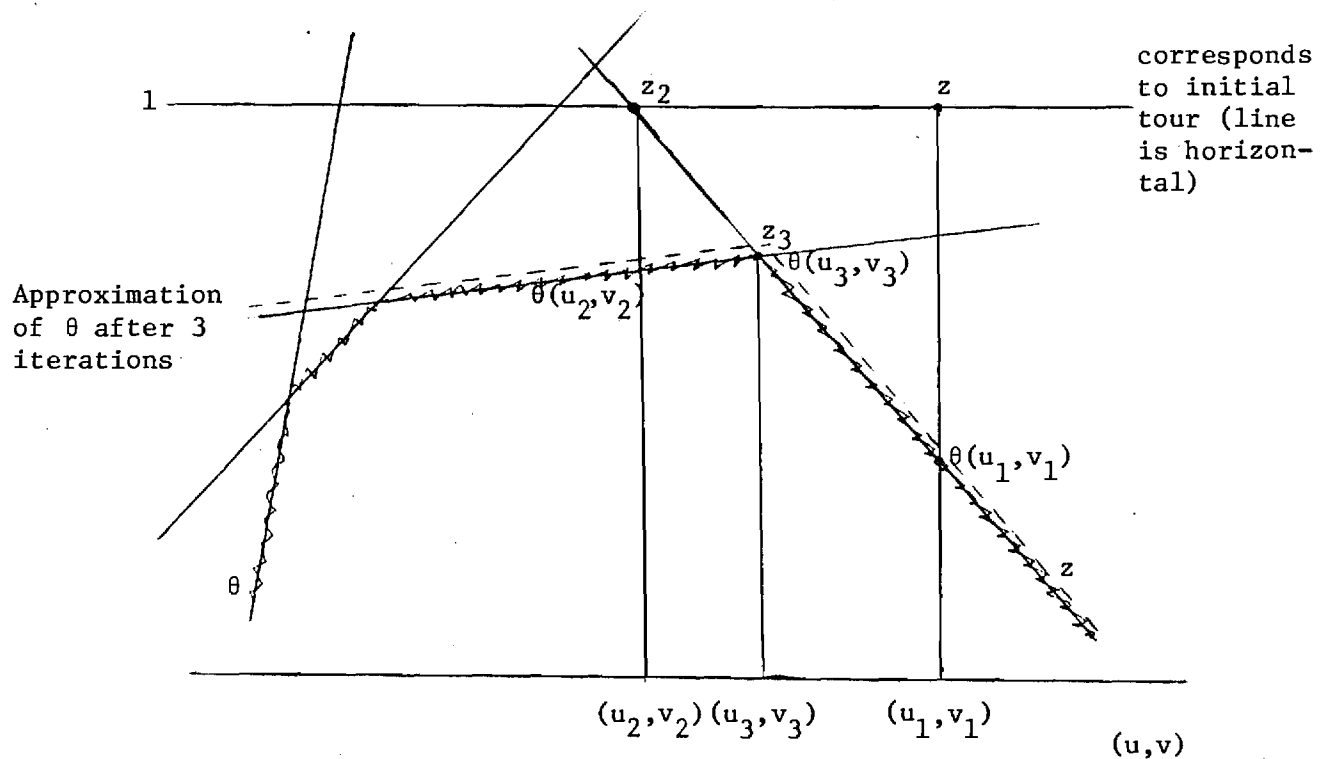
If  $z_k = cx_{k+1} + u_k g(x_{k+1}) + v_k h(x_{k+1})$  then the optimal of the dual problem is  $z_k$  and we stop. Otherwise replace  $k$  by  $k+1$  and go to Step 1.

(Update the column  $\begin{pmatrix} g(x_{k+1}) \\ h(x_{k+1}) \end{pmatrix}$  by multiplying it by the inverse basis of the optimal tableau of Step 1, and put the updated column in the optimal tableau of Step 1.)

Of course, if the optimal 1-tree at any stage is a tour, then we have already obtained the optimal solutions of the primal and dual problems, and we stop. If the "ideal" case is not realized, rather than solving the dual problem exactly, we may terminate whenever the current  $z_k$  is close enough to the best known lower bound up to iteration  $k$ , namely,  $\max_{1 \leq j \leq k} \theta(u_j, z_j)$ .

Finiteness of the above procedure is obvious, since it is precisely the Dantzig-Wolfe decomposition (see, for example, [3,4,14]) applied to the above problem. The above procedure can also be interpreted in many other ways, e.g., as a cutting plane procedure [17], grid optimization [14], and tangential approximation [8,14]. We will interpret the above procedure as a tangential approximation method. This will give a distinct flavor to this strategy, which is significantly different from strategies 2 and 3, which are indeed very similar to each other. Recall that the master problem optimizes an "approximation" of  $\theta$ , namely, the minimum of  $cx_j + u g(x_j) + v g(x_j)$  for  $j = 1, 2, \dots, k$ . So the master problem indeed finds an optimal, not to the overall prob-

lem, but to an approximation of it. Of course, if  $k = t$ , then the approximation is perfect (not necessarily conversely). After that, a decision as to whether the approximation is good enough, that is, whether the optimal to the approximation of  $\theta$  is also the optimal of  $\theta$ , must be made. This is done via the subproblem. The value of  $\theta$  at the optimal dual variables of the approximation problem is evaluated by a simple minimization of a 1-tree problem. If  $z_k \leq cx_{k+1} + u_k g(x_{k+1}) + v_k h(x_{k+1})$  where  $(z_k, u_k, v_k)$  solve the master problem,  $x_{k+1}$  is the optimal of the subproblem; then we must stop. Otherwise, we conclude that the current approximation is not satisfactory, and the new cut,  $z \leq cx_{k+1} + u g(x_{k+1}) + v h(x_{k+1})$  is added to the master problem. This procedure is illustrated below. Note that the search is preferred to start with a complete tour, i.e.,  $g(x_1) = 0$  and  $h(x_1) = 0$ , otherwise the optimal of the first master problem is  $+\infty$ . At any rate, at each iteration, a better approximation of  $\theta$  is realized. At iteration 1,  $z_1$  is the cost of the initial feasible tour. Note that  $z_1 > \theta(u_1, v_1)$ , so the cut  $z \leq cx_1 + u g(x_1) + v h(x_1)$  is added. In the second iteration,  $z_2 > \theta(u_2, v_2)$ , and hence, another cut is added. At iteration 3,  $z_3 = \theta(u_3, v_3)$ , and we stop. Of course, the  $z_j$ 's are monotone decreasing.  $\theta(u_j, v_j)$ 's form lower bounds but are not necessarily monotone. Hence, as the procedure continues, we are not necessarily improving each  $(u_j, v_j)$ , but rather improving the approximation of  $\theta$ , and hence,  $z$ .



### Tangential Approximation of $\theta$

#### Strategy 2 (Steepest Ascent Optimization)

Strategies 2 and 3 improve the value of  $\theta$  until optimality is reached. So, given  $(u, v)$ , we attempt to find a direction such that while moving along it we will increase the value of  $\theta$ . This distinguishes strategies 2 and 3 from strategy 1, which completely optimizes an "approximation" of  $\theta$  at each stage, and hence does not necessarily improve  $\theta$  itself at every iteration. Since  $\theta$  is concave piecewise linear, finding the direction of steepest ascent "looks" promising. Before we discuss in more detail how to find the direction of steepest ascent, we will present precise definitions of ascent and steepest ascent directions.



### Definitions

A vector  $(u,v)$  with norm 1 is called a direction of ascent of  $\theta$  at

$$(u_k, v_k) \text{ if } \theta'((u_k, v_k); (u, v)) = \lim_{\lambda \rightarrow 0^+} \frac{\theta(u_k + \lambda u, v_k + \lambda v) - \theta(u_k, v_k)}{\lambda} > 0.$$

A direction of ascent  $(u^*, v^*)$  is called a direction of steepest ascent

$$\text{if } \theta'((u_k, v_k); (u^*, v^*)) = \text{Maximum}_{\|(u,v)\|=1} \theta'((u_k, v_k); (u, v)).$$

If  $\theta$  has no directions of ascent at  $(u_k, v_k)$ , then the steepest ascent

at  $(u_k, v_k)$  is called zero. By concavity of  $\theta$ , it is clear that  $\theta$  attains its maximum at  $(u_k, v_k)$  if the steepest ascent there is zero.

### Characterization of Ascent and Steepest Ascent Directions

Suppose that  $(u_k, v_k)$  is given, with  $\theta(u_k, v_k) = cx_j + u_k g(x_j) + v_k h(x_j)$

$$\text{for } j \in I. \text{ Note that } \theta'((u_k, v_k); (u, v)) = \lim_{\lambda \rightarrow 0^+} \frac{\theta(u_k + \lambda u, v_k + \lambda v) - \theta(u_k, v_k)}{\lambda}$$

$$= \text{Minimum}_{\zeta \in E} \langle \zeta, (u, v) \rangle, \text{ where } E \text{ is the collection of subgradients of}$$

$\theta$  at  $(u_k, v_k)$ . (See, for example, Rockafellar [15] and Grinold [9]).

But  $\zeta \in E$  can be characterized as:

$$\sum_{j \in I} \lambda_j \begin{pmatrix} g(x_j) \\ h(x_j) \end{pmatrix}$$

$$\sum_{j \in I} \lambda_j = 1$$

$$\lambda_j \geq 0 \quad \text{for } j \in I.$$

Therefore,  $(u, v)$  is an ascent direction if and only if  $u g(x_j) + v h(x_j)$

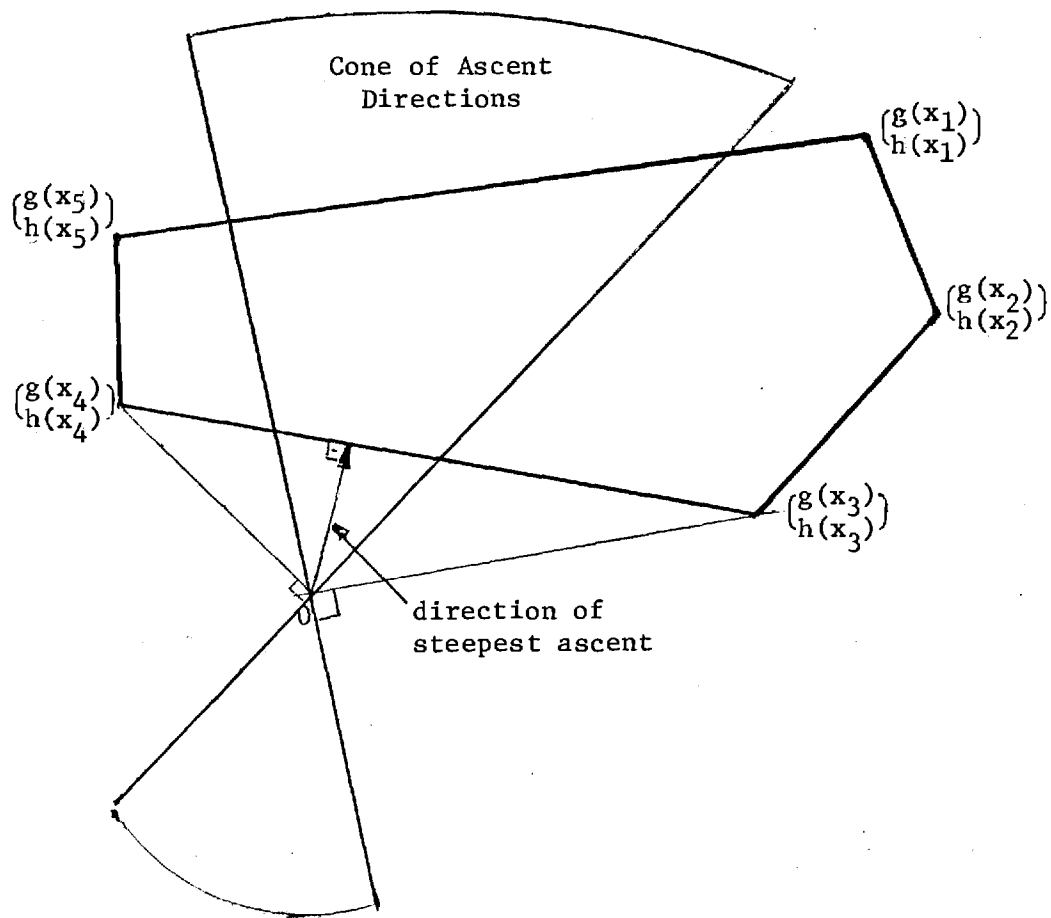
$> 0$  for each  $j \in I$ . A steepest ascent direction can further be characterized as follows:

$$\begin{aligned} \theta'((u_k, v_k); (u^*, v^*)) &= \text{Maximum}_{\|(u,v)\|=1} \theta'((u_k, v_k); (u, v)) \\ &= \text{Maximum}_{\|(u,v)\|=1} \text{Minimum}_{\zeta \in E} \langle \zeta, (u, v) \rangle \\ &= \text{Minimum}_{\zeta \in E} \text{Maximum}_{\|(u,v)\|=1} \langle \zeta, (u, v) \rangle \end{aligned}$$

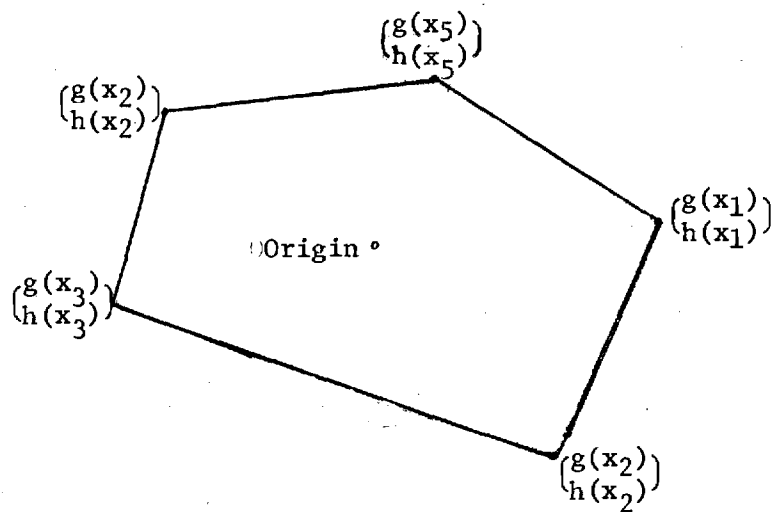
where the Max Min is replaced by Min Max due to linearity in  $\zeta$  and  $(u,v)$  and compactness of  $E$  and the set  $\{(u,v) : \|(u,v)\| = 1\}$ . (See, for example, [1].) If  $\zeta \neq 0$  then the Maximum  $\langle \zeta, (u,v) \rangle$  is given by letting  $(u,v) = \frac{\zeta}{\|\zeta\|}$ . Hence,  $\theta'((u_k, v_k); (u^*, v^*)) = \text{Minimum}_{\zeta \in E} \|\zeta\|$ . In particular, the direction of steepest ascent is then the subgradient with minimum length, normalized by dividing by its norm.

### Geometric Interpretation

Given a point  $(u_k, v_k)$  where  $\theta(u_k, v_k) = cx_j + u_k g(x_j) + v_k h(x_j)$  for  $j \in I$ , there is an interesting geometric interpretation of ascent and steepest ascent directions. Consider the lattice points  $\begin{pmatrix} g(x_j) \\ h(x_j) \end{pmatrix}$  for  $j \in I$  in the  $(u,v)$  space. The convex combinations of these points is a polyhedron and forms the set of subgradients of  $\theta$  at  $(u_k, v_k)$ . The lattice points  $\begin{pmatrix} g(x_j) \\ h(x_j) \end{pmatrix}$  for  $j \in I$  are the extreme points of this polyhedron. Not all points in the polyhedron are ascent directions.  $(u,v)$  is an ascent direction if and only if  $ug(x_j) + vh(x_j) > 0$  for each  $j \in I$ , that is, if it belongs to the cone shown below ( $I = \{1, 2, 3, 4, 5\}$ ). In particular, the steepest ascent direction is the subgradient (normalized so that it has norm 1) closest to the origin. In particular, if the polyhedron contains the origin, then the steepest ascent is zero, and  $(u_k, v_k)$  is the maximum of  $\theta$ . The process of finding the steepest ascent has actually been used in many contexts. The above development gives a geometric characterization of these directions. In fact, the infeasibility decomposition procedure of Balas [2], the procedure of Grinold [9], and more recently the procedure of Fisher, Northup, and Shapiro [6] can be viewed of methods for finding the steepest ascent



Directions of Ascent and Steepest Ascent

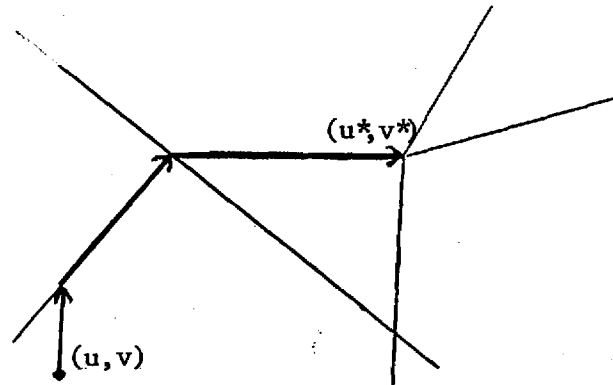


Steepest Ascent is Zero

direction with the  $\ell_1$ -norm so that linear programming can be used.

In Bazaraa and Goode [1], a procedure for finding the steepest ascent with Euclidean norm is presented.

We will describe some of the advantages and disadvantages of moving along the steepest ascent direction. The advantage of the steepest ascent optimization is obvious. We move along the best local optimizing direction, either across regions, or along ridges, until the optimal is reached in a finite number of steps, as illustrated below.



Moving Along Steepest Ascent Directions

The disadvantage of a steepest ascent procedure is that finding the steepest ascent direction, if we are in the intersection of many regions, is not an easy task. Before the shortest convex combination of  $\{g(x_j)\}_{j \in I}$  is found, we must find  $\{g(x_j)\}_{j \in I}$ , either at once, or through a generic step, which may not be a trivial matter in this case.

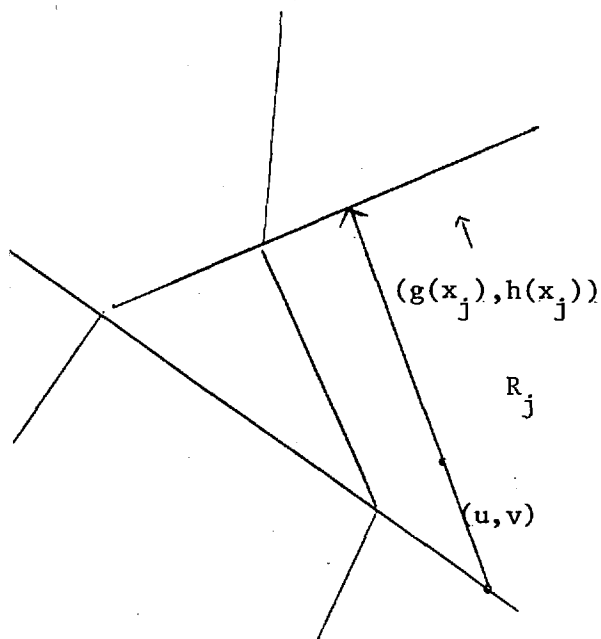
Finally, even if we find the direction of steepest ascent, finding how far to move along it, is not trivial, since we have to find the point

at which a new 1-tree becomes an alternative optimal (an LP based calculation may prove useful here. Of course, an approximation through utilization of discrete steps may be used, as in Held and Karp [11] (also see [12]) in their subgradient optimization method of the symmetric traveling salesman problem.

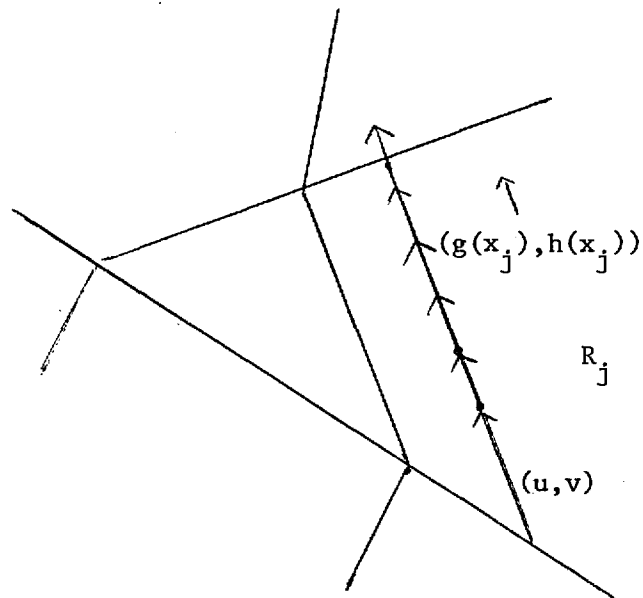
### Strategy 3 (Subgradient Optimization)

Rather than moving along steepest ascent directions, one can move along subgradients, and in particular, along extreme points of the polyhedron formed by the subgradients, namely,  $\begin{pmatrix} g(x_j) \\ h(x_j) \end{pmatrix}$  for  $j \in I$ . This procedure has been successfully used by Held and Karp [11].

Of course, if we are at a point  $(u,v)$  which belongs to the interior of a region  $R_j$ , i.e., if there exists a unique minimal 1-tree where the cost of link  $ij$  is  $c_{ij} + u_i + v_j$ , then  $\theta$  is differentiable at  $(u,v)$  and the unique subgradient is  $\begin{pmatrix} g(x_j) \\ h(x_j) \end{pmatrix}$ , and moving along it a positive distance will increase  $\theta$ . Of course, the question is how far to move along this direction. Needless to say,  $\theta((u,v) + \lambda(g(x_j), h(x_j)))$  is an increasing function of  $\lambda$  as long as  $(u,v) + \lambda(g(x_j), h(x_j)) \in R_j$ . Rather than finding the value of  $\lambda$  at which  $(u,v) + \lambda(g(x_j), h(x_j))$  hits the boundary of  $R_j$ , i.e., when an alternative minimal 1-tree appears, one may move a small discrete step along  $(g(x_j), h(x_j))$ , and then calculate  $\theta$  at the new point. If the minimal 1-tree is still  $x_j$ , then we still move a discrete step in the direction  $(g(x_j), h(x_j))$  and repeat the process. Here there is a tradeoff between the computational effort spent in finding how far to move before leaving  $R_j$ , and the computational effort in repeating the minimal 1-tree calculation several times, one every time a discrete step is taken.



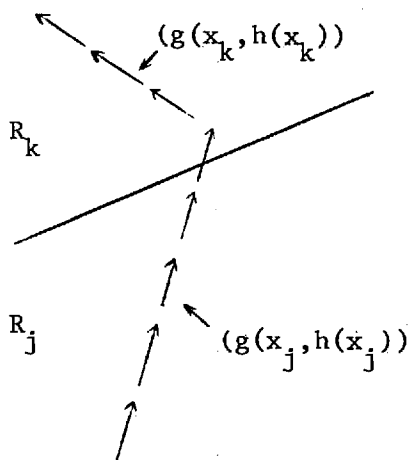
Moving until the Boundary  
of  $R_j$  is reached



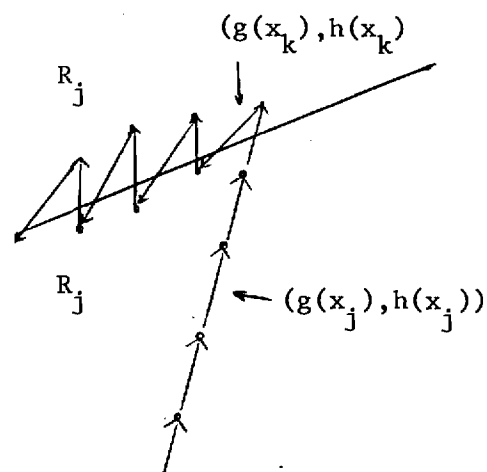
Moving in Discrete Steps

As long as we move along  $(g(x_j), h(x_j))$  and stay in region  $R_j$ ,  $\theta$  is monotonically increasing. However, when we "step into" a new region,  $\theta$  may or may not increase, since the gradient of the new region may point back into the region we just came from. At any rate, even if the last step did not incur any improvement, recalculating the minimal  $l$ -tree at the new point will lead to a direction which improves  $\theta$ , provided of course, we are in the interior of the new region. From this discussion it is clear that this type of discrete search is really a gradient search, rather than a subgradient search, because the likelihood of being exactly in the intersection of more than one region is small, and  $(g(x_j), h(x_j))$  is indeed the gradient of  $\theta$  provided that we are in the interior of  $R_j$ , i.e., when  $x_j$  is the unique minimal  $l$ -tree. The two cases illustrated below are possible. In the first case, moving along the new gradient tends to take us into the interior of the new region, and hence we can move an appreciable distance across  $R_k$ , thus improving  $\theta$ . The other case shows a situation where moving along

$(g(x_k), h(x_k))$  will take us back to  $R_j$ , i.e.,  $\langle (g(x_k), h(x_k)), (g(x_j), h(x_j)) \rangle < 0$ . This case corresponds to  $\theta$  improving along the ridge  $R_j \cap R_k$ , and the search procedure described above will lead to zigzagging as shown below.



Moving Along New Gradient  
Takes us "Away from  $R_j$ "

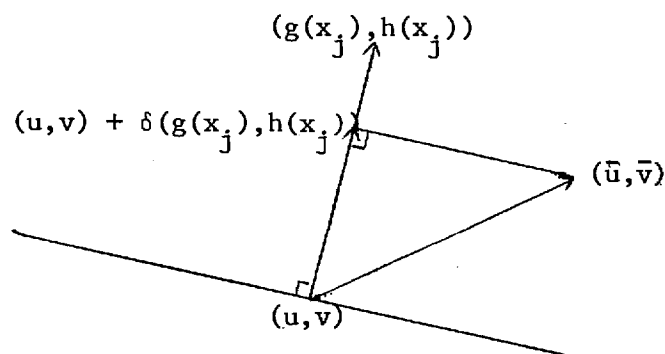


Moving Along New Gradient  
Takes us Back to  $R_j$  "zigzagging"

It is also interesting to note, that even if moving along the direction  $(g(x_j), h(x_j))$  (or any other subgradient of  $\theta$  at  $(u, v)$ ) starting from  $(u, v)$  does not improve the objective function, it will at least get us closer to the optimal point, provided that the stepsize is small enough. This fact has been shown by Held and Karp [11]. We will now illustrate this fact geometrically. Note that the half space  $\{(\hat{u}, \hat{v}) : \langle (\hat{u}, \hat{v}) - (u, v), (g(x_j), h(x_j)) \rangle \geq 0\}$  contains the optimal point  $(\bar{u}, \bar{v})$ . This is obvious since by concavity of  $\theta$  we have

$$\theta(\bar{u}, \bar{v}) \leq \theta(u, v) + \langle (\bar{u}, \bar{v}) - (u, v), (g(x_j), h(x_j)) \rangle$$

Since  $\theta(\bar{u}, \bar{v}) \geq \theta(u, v)$  then  $\langle (\bar{u}, \bar{v}) - (u, v), (g(x_j), h(x_j)) \rangle \geq 0$ . From this it is then clear that moving along the normal vector  $(g(x_j), h(x_j))$  to the hyperplane  $\{(\hat{u}, \hat{v}): \langle (\hat{u}, \hat{v}) - (u, v), (g(x_j), h(x_j)) \rangle = 0\}$  starting from  $(u, v)$  leads closer to  $(\bar{u}, \bar{v})$ . In fact  $\|(\bar{u}, \bar{v}) - (u, v) - \lambda(g(x_j), h(x_j))\|$  is a monotone decreasing function of  $\lambda$  for  $\lambda \in [0, \delta]$  where  $\delta$  is found such that the vectors  $(g(x_j), h(x_j))$  and  $(\bar{u}, \bar{v}) - (u, v) - \delta(g(x_j), h(x_j))$  are orthogonal (see Figure below), i.e.,  $\delta = \frac{\langle (g(x_j), h(x_j)), (\bar{u}, \bar{v}) - (u, v) \rangle}{\|(g(x_j), h(x_j))\|^2}$ .



How Far to Go?

Even though the above analysis will not immediately lead to finding a suitable stepsize (since  $(\bar{u}, \bar{v})$  is not known at this stage), it at least establishes the fact that even if the objective function does not improve by moving along  $(g(x_j), h(x_j))$  (if  $(u, v)$  is in the intersection of more than one region), we at least get closer to the optimal point, provided that the stepsize is small enough. As we get closer to the optimal point  $(\bar{u}, \bar{v})$ , the vector  $(\bar{u}, \bar{v}) - (u, v)$  becomes small in norm and noting the form of  $\delta$  it becomes clear that the stepsize should be small.



### Zigzagging and Averaging of Directions

The case of zigzagging along the ridge  $R_j \cap R_k$  may be time consuming. An averaging, or taking a suitable convex combination of the directions at any two successive iterations, will help reduce the zigzagging. This will in effect act as a projection of either of the gradient vectors on  $R_j \cap R_k$ . Calculating the projection vector along  $R_j \cap R_k$  is easy, but the projection in the intersection of several regions may be tedious. Of course, if we are moving across the interior of a region  $R_j$ , then averaging of the gradient vector  $(g(x_j), h(x_j))$  and itself will result in  $(g(x_j), h(x_j))$ , i.e., no harm is done!

### Stepsize

Suppose we are at the point  $(u,v)$  with minimal 1-tree  $x_j$ . If  $(u,v)$  is in the interior of  $R_j$ , which is the case quite often, then moving along  $(g(x_j), h(x_j))$  a small distance such that we are still in region  $R_j$  will improve the objective function. If we are very close to the boundary of  $R_j$  or on it and move along  $(g(x_j), h(x_j))$  with a stepsize equal to  $\lambda$ , the objective function may or may not improve. Even if it does not improve, we get closer to the optimal point, provided that  $\lambda$  is small enough. At any rate, in order to implement the above procedure, we have to choose a suitable stepsize (or stepsizes). Normalization of the subgradient vector  $(g(x_j), h(x_j))$  will be helpful so that the stepsize becomes more meaningful. According to our computational experience, we found that the stepsize does not play a major role in affecting the computational time for the traveling salesman problem, especially since we are seeking "close to optimal" solu-

tions of the dual problem rather than true optimal solutions. Discrepancy between suboptimal dual solutions and the optimal dual solution, as well as any duality gap, will be compensated by the branch and bound scheme. We found that a value of  $\lambda = 1$  is quite satisfactory. As a matter of fact, a value of  $\lambda$  anywhere from 1 to 5 was found satisfactory for all the problems we tried. A combination of different strategies can also be used, starting with a larger  $\lambda$  and then switching to a smaller  $\lambda$  as we get no improvement in the value of  $\theta$  for several iterations.

#### Subgradient Optimization Algorithm

We will now describe a simple subgradient algorithm for maximizing  $\theta$ . As in the previously discussed strategies we may not reach the true optimal of the dual function but rather get sufficiently close to it. Convergence of the procedure with sufficiently small  $\lambda$  is shown in [11]. We will use the a fixed value of  $\lambda$  (say  $\lambda = 1$ ) in the following description with the understanding that different monotone values as the procedure continues can be used. At each iteration a minimal 1-tree is calculated. Different choices of the initial node can be made. Here  $n$  problems are initially solved with the initial dual variables  $u_1 = v_1 = 0$ , each time finding the minimal 1-tree with a different initial node. The initial node throughout the algorithm is the one with maximum minimal 1-tree.

#### Initialization Step

Choose  $(u_1, v_1)$ , say  $u_1 = v_1 = 0$ , or  $u_1$  and  $v_1$  the dual variables of the corresponding assignment problem. Let  $u^* = u_1$  and  $v^* = v_1$ . Let

$k = 1$ ,  $\theta^* = -\infty$ ,  $\gamma = 0$ , and  $v_{\max} = T$ . ( $T$  is the maximum number of iterations allowed without improving  $\theta$ .) Choose  $\lambda$  and go to Step 1.

Step 1 (Finding  $\theta(u_k, v_k)$ )

Solve the following minimal 1-tree subproblem.

$$\text{Minimize} \quad \sum_{\substack{j=1 \\ j \neq i}}^n \sum_{i=1}^n (c_{ij} + u_{ik} + v_{jk}) x_{ij}$$

Subject to:  $x \in X$

where  $u_k = (u_{1k}, \dots, u_{nk})$  and  $v_k = (v_{1k}, \dots, v_{nk})$

Let  $x_k$  be an optimal 1-tree. If  $x_k$  is a directed loop, then stop, the optimal primal and dual solutions have been found. Otherwise

$$\text{let } \theta(u_k, v_k) = \sum_{\substack{j=1 \\ j \neq i}}^n \sum_{i=1}^n (c_{ij} + u_{ik} + v_{jk}) x_{ij} - \sum_{i=1}^n u_{ik} - \sum_{j=1}^n v_{jk}.$$

If  $\theta(u_k, v_k) > \theta^*$ , then let  $v = 0$ , replace  $\theta^*$  by  $\theta(u_k, v_k)$ , replace  $u^*$  by  $u_k$  and  $v^*$  by  $v_k$ , and go to Step 2. If on the other hand,  $\theta(u_k, v_k) \leq \theta^*$ , replace  $v$  by  $v+1$ . If  $v = v_{\max}$ , then stop with  $\theta^*$  and  $(u^*, v^*)$ . If  $v < v_{\max}$ , go to Step 2.

Step 2 (Update dual variables)

Let  $u_{k+1} = u_k + \lambda g(x_k)$  and  $v_{k+1} = v_k + \lambda h(x_k)$ . Replace  $k$  by  $k+1$  and repeat Step 1.

Finding the Optimal Primal Solution: Branch and Bound

So far we have described various strategies for finding the optimal (or near optimal if ascent search is not fully completed) dual solution. In the process, the optimal loop of the traveling salesman problem may be found. If this were not the case (either a duality gap exists or

dual is not solved exactly), a branch and bound scheme for finding the optimal directed loop will be described. It is important to note that even in this case, optimizing the dual function  $\theta$  played the following significant roles:

1. Provided a "tight" lower bound on the optimal traveling salesman problem and hence making the notion of "verified" suboptimal solutions attractive.
2. Rather than using the cost  $c_{ij}$  of link  $ij$  in the branch and bound scheme,  $c_{ij}$  is replaced by  $c_{ij} + u_i + v_j$ , where  $u$  and  $v$  are the optimal dual variables, thus resulting in a "well-conditioned" cost matrix. The computational effort for finding the optimal tour, using the new cost matrix  $c_{ij} + u_i + v_j$  is considerably smaller than the effort to be expended, had the original cost matrix been used in a branch and bound scheme. This is due to the fact that the new cost matrix  $(c_{ij} + u_i + v_j)$  is "conditioned" in the sense that the optimal 1-tree with the above matrix is "close" to the optimal traveling salesman tour.

Consider now the following traveling salesman problem:

$$\begin{array}{ll}
 \text{Minimize} & \sum_{\substack{j=1 \\ j \neq i}}^n \sum_{i=1}^n (c_{ij} + u_i + v_j) x_{ij} \\
 \text{Subject to:} & \sum_{\substack{j=1 \\ j \neq i}}^n x_{ij} = 1 \quad \text{for } i = 1, 2, \dots, n \\
 & \sum_{\substack{i=1 \\ i \neq j}}^n x_{ij} = 1 \quad \text{for } j = 1, 2, \dots, n \\
 & \text{No Subtours}
 \end{array}$$

Note that an optimal tour obtained by the above problem is an optimal solution of the original traveling salesman problem. A branch and bound procedure will be used to find the optimal tour.

#### General Framework

A branch and bound procedure will be used to obtain the optimal tour. A list of distinct nodes  $i_1, i_2, i_3, \dots, i_k$  and corresponding links  $(i_1, i_2), \dots, (i_{k-1}, i_k)$  form a partial solution. Since the nodes are distinct, no subtours are a part of the partial solution. If  $k=n$ , then a complete tour is obtained by adding the link  $(i_k, i_1)$ . Suppose that we are given a partial solution, and attempt to find a feasible completion, i.e., a tour. A lower bound on the cost of completing this partial solution is computed, say LB. If  $LB \geq C^*$ , where  $C^*$  is the cost of the best tour known so far, then the partial solution is fathomed, since there is no way to find a better solution by completing it. In this case the partial solution is abandoned. If  $LB < C^*$  then an additional link  $(i_k, i_{k+1})$  is added to the partial tour where  $i_{k+1}$  is not equal to  $i_j$  for  $j = 1, 2, \dots, k$ . The method of choosing  $i_{k+1}$ , referred to as the branching node, and the method of calculating the lower bound, completely define the search.

#### Calculation of the Lower Bound

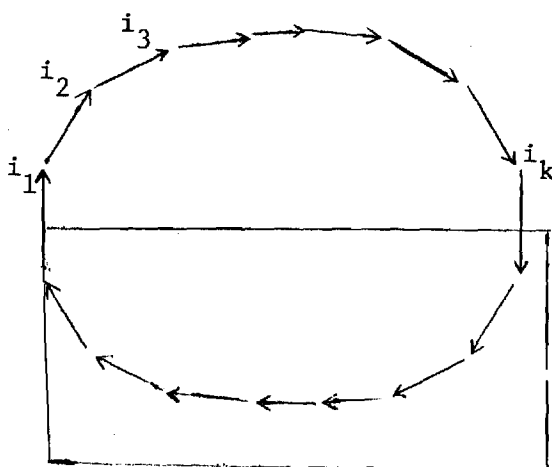
Suppose that  $i_1, i_2, \dots, i_k$  form a partial solution. If  $(i_j, i_{j+1})$  for  $j = 1, 2, \dots, k-1$  are to be a part of a tour, then the remaining nodes must themselves form a "string", which is connected to nodes  $i_1$  and  $i_k$ . Since the "string" is a tree spanning the remaining nodes, then its cost must be greater or equal to the minimal cost of

all trees spanning these nodes. Also, the cost of connecting the partial solution to this open loop is greater or equal to the sum of the following:

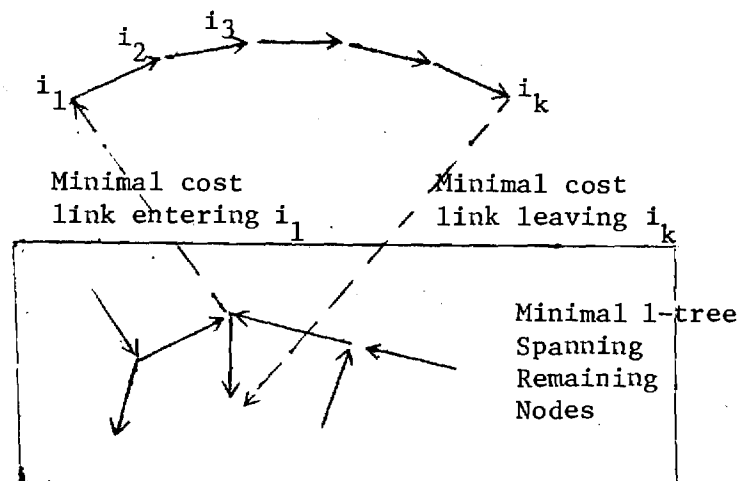
1. Minimal cost of (available) links leaving node  $i_k$
2. Minimal cost of (available) links entering node  $i_1$

We use the word available, since links  $(i_j, i_1)$  for  $j = 2, \dots, k$  and  $(i_k, i_j)$  for  $j = 1, 2, \dots, k-1$  are not available, because their use will form a subtour. Also, as the search continues, other links will be prohibited from usage.

We show below a feasible completion of a partial solution, and also a method for calculating a lower bound on the cost of all completions of this partial solution.



A Feasible Completion  
of the Partial Solution



Calculation of Lower Bound on All  
Completions of Partial Solution

From this it is clear that a lower bound LB on the cost of the partial solution, plus the cost of its completion, is given by:

$$LB = C_1 + C_2 + C_3, \text{ where}$$

$C_1$  : the actual cost of the partial solution

$$C_1 = \sum_{j=1}^{k-1} (c_{i_j i_{j+1}} + u_{i_j} + v_{i_{j+1}})$$

$C_2$  : the cost of the minimal tree spanning the remaining nodes, where

the cost of link  $ij$  is  $c_{ij} + u_i + v_j$ . There are efficient methods for calculating  $C_2$  (see [11]).

$C_3$  : Minimal cost of "connecting" the partial solution to the spanning tree.

$$C_3 = \text{Minimum}_{j \in A} (c_{i_k j} + u_{i_k} + v_j) + \text{Minimum}_{j \in B} (c_{j i_1} + u_j + v_{i_1})$$

where  $A$  is the set of nodes which  $i_k$  is eligible to enter, and

$B$  is the set of nodes which are eligible to enter node  $i_1$ .

### Continuation of the Search

#### 1. Fathoming (Backward Move)

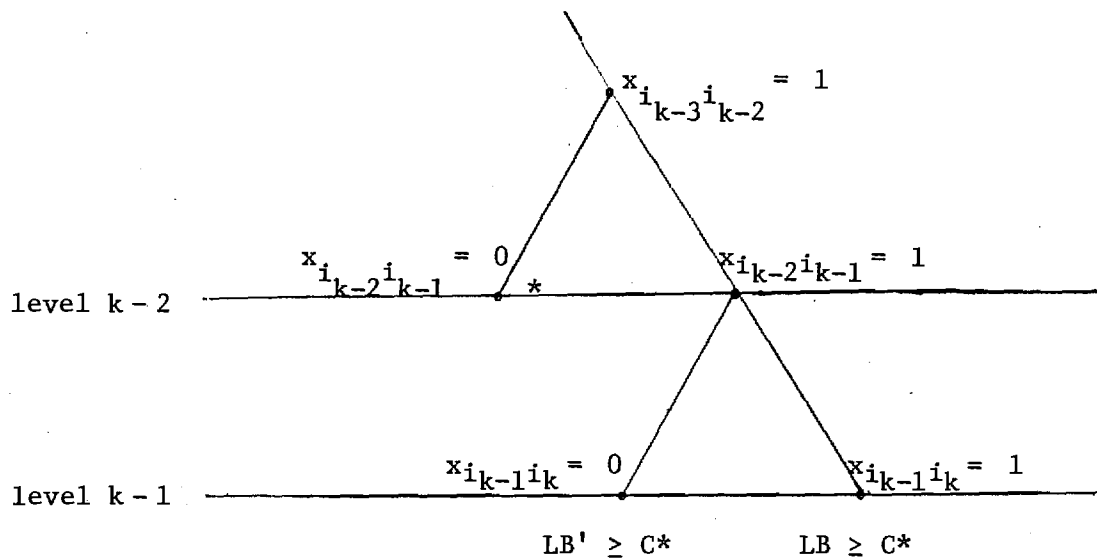
Suppose that we have a partial tour  $i_1, i_2, \dots, i_k$ ; i.e.,  $x_{i_1 i_2} = x_{i_2 i_3} = \dots = x_{i_{k-1} i_k} = 1$ . The level of the search tree is called  $k-1$ .

The lower bound LB on the cost of the partial solution plus its completion is calculated as shown above. If  $LB \geq C^*$ , where  $C^*$  is the current best known cost of a complete tour, then pursuing this partial solution is not worthwhile, and the partial solution is said to be fathomed. The last move from  $i_{k-1}$  to  $i_k$  is prohibited, i.e.,

$x_{i_{k-1} i_k} = 0$ . A new bound  $LB'$  on completing the cost of the partial tour  $i_1, \dots, i_{k-1}$  with the additional restriction that  $x_{i_{k-1} i_k} = 0$  is computed. This is done in the same manner as before except of

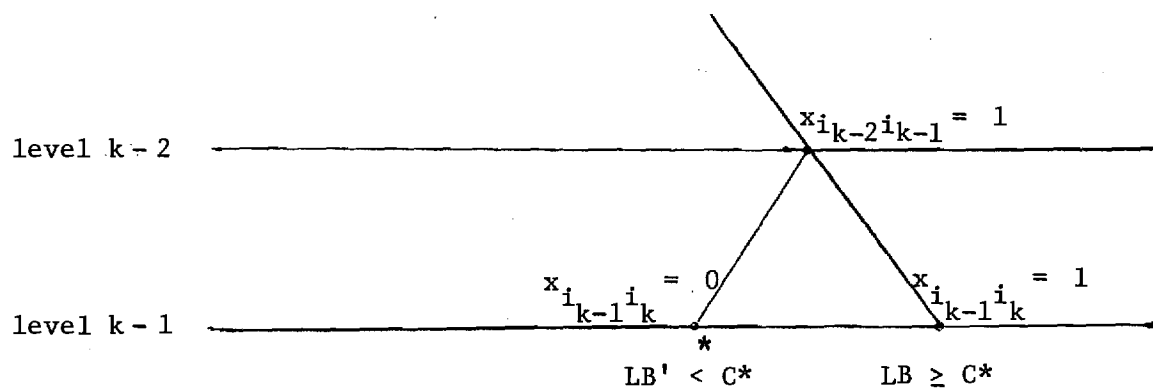
course that  $i_k$  now belongs to the remaining nodes and the link  $i_{k-1}i_k$  cannot be used. If  $LB' \geq C^*$ , then the partial solution  $i_1, i_2, \dots, i_{k-1}$ , with the added restriction  $x_{i_{k-1}i_k} = 0$  can lead to no improved solutions. This means that all possibilities at level  $k-1$  are exhausted. This case will be called a strong fathoming, and the level of the tree is reduced by 1 unit. In this case  $x_{i_{k-1}i_k} = 0$  is not prohibited any more, whereas  $x_{i_{k-2}i_{k-1}}$  is forced to be zero. The process is then repeated. If on the other hand  $LB' < C^*$ , which will be called weak fathoming, then we seek to find a node  $j$  such that  $x_{i_{k-1}j} = 1$ . This step will be discussed in more detail in the forward move below. The cases of strong and weak fathoming are illustrated below.





#### Strong Fathoming

\* Continue at level  $k-2$  by trying to go from node  $i_{k-2}$  to a node  $j \neq i_{k-1}$



#### Weak Fathoming

\* Continue at level  $k-1$  by trying to go from node  $i_{k-1}$  to node  $j \neq i_k$ .

## 2. Forward Move

If the lower bound is less than  $C^*$ , then it may be possible to find a completion of the partial solution with an objective less than  $C^*$ .

Suppose that the current level of the tree is  $k-1$ , where at level  $k-1$  we have  $x_{i_{k-1}i_k} = 1$ . Now we seek a node  $i_{k+1} \neq i_j$  for  $j = 1, 2, \dots, k-1$  and the move  $i_k i_{k+1}$  is not prohibited from previous fathoming (if such a node  $i_{k+1}$  is not found then the partial solution is fathomed and the level of the tree is reduced). Many rules (branching rule) for choosing  $i_{k+1}$  are possible. One of the simplest (and most effective for the traveling salesman problem) is to choose  $i_{k+1}$  which minimizes  $c_{i_k j} + u_{i_k} + v_j$  for all available  $j$ . Another possible method is to calculate a lower bound  $LB_j$  for each possible node  $j$ , and then choose the one with the least possible lower bound to be  $i_{k+1}$ . If the choice of  $i_{k+1}$  is determined by this rule, then we automatically have the lower bound needed for deciding whether to fathom, and we also may prohibit some of the moves  $i_k j$  if  $LB_j \geq C^*$ .

At any rate,  $i_{k+1}$  is chosen, the level of the tree is increased by 1, and a lower bound (if the first rule was used) is calculated, and the process is repeated. Needless to say, if the level is  $n-1$ , then  $i_n$  is joined with  $i_1$  and the cost of the link  $i_n i_1$  is added to the cost of the partial tour. If the total cost is less than  $C^*$  then  $C^*$  is replaced by the total cost, and the tour is stored as the best tour obtained so far.

### 3. Termination

We have described forward and backward progress of the search. If the level of the search ever reaches value zero, then we stop. This would mean that we are currently at level 1, and are trying to back-track. This means that all possible completions of  $x_{i_1 i_2} = 1$  and  $x_{i_1 i_2} = 0$  have been enumerated, i.e., all the search space has been exhausted. The stored loop and  $C^*$  give the optimal solution.

### Summary of the Algorithm

#### Initialization Step

Let  $P(i) = \emptyset$  for all  $i = 1, 2, \dots, n$  and let  $C^* = \infty$ . Choose node  $i_1$  (according to the procedure described earlier) as the initial node in the maximization of  $\theta$ . Let  $k = 0$  and go to Step 1.

#### Step 1 (Forward Move)

If  $k = n - 1$ , replace  $C_1$  by  $C_1 + (c_{i_n i_1} + u_{i_n} + v_{i_1})$ . If  $C_1 < C^*$ , then replace  $C^*$  by  $C_1$ , store the complete solution  $i_1, \dots, i_n, i_1$  and go to Step 2. If  $C_1 \geq C^*$ , go to Step 2. If  $k < n - 1$  calculate a lower bound LB on all completions of the partial solution at hand, where  $LB = C_1 + C_2 + C_3$ , where  $C_1, C_2$ , and  $C_3$  are calculated as described above,  $A = P(i_k) \cup \{i_1, \dots, i_{k-1}, i_k\}$  and  $B = \{i_1, i_2, \dots, i_k\}$ . If  $LB \geq C^*$ , go to Step 2. Otherwise pick  $i_{k+1} \notin A$  such that  $c_{i_k i_{k+1}} + u_{i_k} + v_{i_{k+1}} = \text{Minimum}_{j \notin A} (c_{i_k j} + u_{i_k} + v_j)$ . Replace  $k$  by  $k+1$  and repeat Step 1.

#### Step 2 (Fathom)

Replace  $P(i_{k-1})$  by  $P(i_{k-1}) \cup \{i_k\}$ . Calculate a lower bound LB on all completions of the partial solution  $i_1, i_2, \dots, i_{k-2}, i_{k-1}$ . If  $LB \geq C^*$ ,

go to Step 3. Otherwise choose a node  $j \notin P(i_{k-1})$  such that

$c_{i_{k-1}j} + u_{i_{k-1}} + v_j = \text{Minimum } \{ c_{i_{k-1}i} + u_{i_{k-1}} + v_i; i \notin P(i_{k-1}), i \neq i_1, \dots, i \neq i_{k-1} \}$ . Denote the new  $i_k$  by  $j$ . This forms a new partial solution. Go to Step 1.

### Step 3 (Strong Fathoming)

$i_{k-1}$  is placed in  $P(i_{k-2})$ .  $P(i_{k-1})$  is replaced by  $\emptyset$ , and  $k$  is replaced by  $k-1$ . If  $k = 0$ , go to Step 4, otherwise go to Step 2.

### Step 4 (Termination)

Search has been completed. Optimal cost is  $C^*$  and its corresponding tour is  $i_1, i_2, \dots, i_n, i_1$ . Stop.

### Optimal and Suboptimal Solutions via "Controlled" Fathoming

Rather than attempting to fathom a partial solution when the lower bound LB on its completion is greater than or equal to the best known  $C^*$ , we may fathom if LB is "Close" to  $C^*$ . This will enable us to fathom quickly and hence reduce the portion of the search tree explicitly enumerated. This procedure can be controlled either to obtain suboptimal solutions, with any desired degree of accuracy, or to obtain optimal solutions. In both cases the computational effort will be significantly reduced. The following two methods are proposed for implementing the notion of controlled fathoming.

#### Method 1

Recall that a partial solution is fathomed, if the lower bound on all completions is at least as big as  $C^*$ , the best known objective. Suppose that a partial solution is fathomed if  $LB \geq \alpha C^*$ , where  $\alpha \in (0,1]$ . In

this case, the partial solution is abandoned if there is no hope that it leads to an objective which is better than  $\alpha C^*$ . To illustrate, suppose that  $C^*$  is 1000, and we choose  $\alpha = 0.95$ . Then we fathom if  $LB \geq 950$ . This means that if the bound on completions of the partial solution cannot lead to an improvement of 50 units over what we already have, then it does not pay off to pursue this partial solution. The objective of this simple strategy is clear. We want to fathom, i.e., abandon partial solutions, quickly, even if they lead to a slight improvement. As the reported computational experience shows, this simple strategy speeds the computational effort considerably. Of course, as a new  $C^*$  is found, then we fathom whenever the bound is greater or equal to  $\alpha$  times the new  $C^*$ . The procedure continues until we cannot find a feasible solution with an objective less than  $\alpha C^*$ . So we have a feasible tour with objective  $C^*$ , coupled with the statement that the optimal objective is greater than or equal to  $\alpha C^*$ . Needless to say, if  $\alpha C^*$  is less than the overall lower bound, which is  $\text{Max } \theta(u,v)$ , then there is no need to complete the search, since we know that there are no solutions with objective less than  $\alpha C^*$ . It is interesting to keep in mind that when we try to find a solution with objective less than  $\alpha C^*$ , the computational effort will either lead to such a solution or conclude that no such solution exists. In the first case, we are able to improve the current best solution, and in the second case we are able to "tighten" the overall lower bound on the problem. In other words, a valuable information is obtained whether we are able to find a solution with objective less than  $\alpha C^*$  or not.

### Choice of $\alpha$

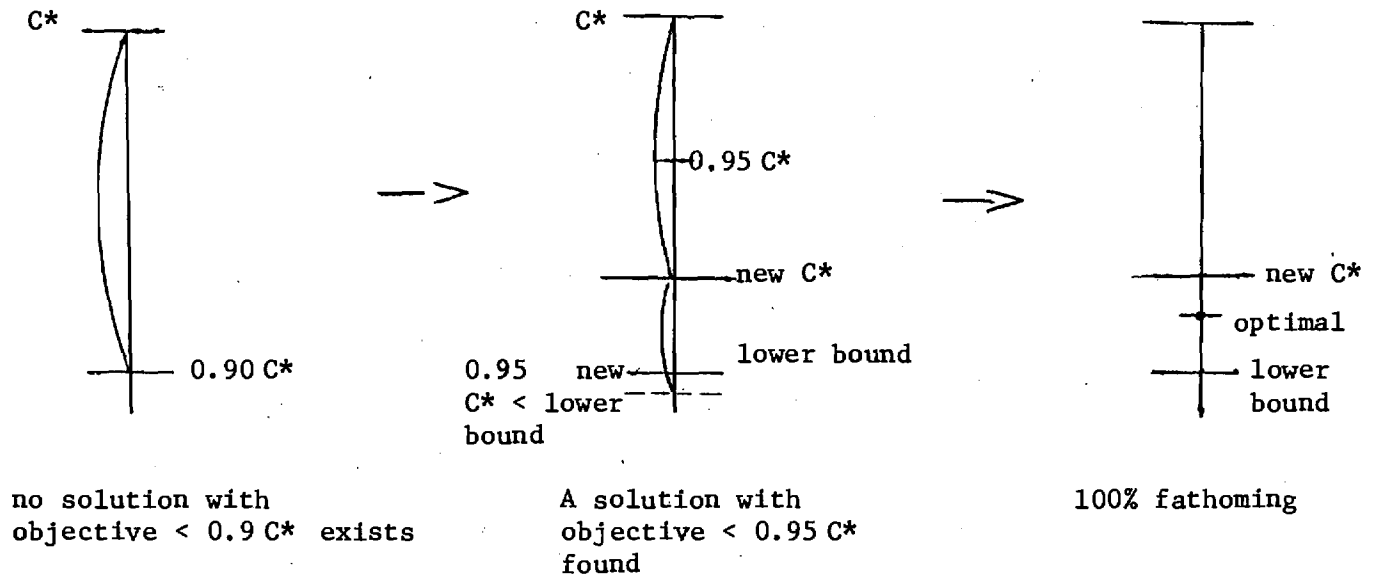
Of course, if  $\alpha$  is small, then fathoming will speed up considerably, resulting in a small computational effort. But on the other hand, the quality of the best obtained feasible solution is not guaranteed. To illustrate, suppose that at termination  $C^*$  is 1000 and suppose that  $\alpha = 0.60$ . All we can say in this case is that we have a feasible solution with objective 1000, and the true optimal has an objective greater or equal to 600. This, of course, is not satisfactory. Suppose, however, that we used  $\alpha = 0.95$ . Then if at termination  $C^*$  is 1000, we can say that the optimal is greater than or equal to 950. This statement is indeed much stronger than the previous statement.

From this discussion, it is clear that there is a tradeoff between the computational effort and the value of  $\alpha$ . We recommend values of  $\alpha \geq 0.93$  depending on the accuracy required.

### Exact Solution By Stepped Fathoming

A slight modification of the above scheme can be used to find the exact optimal solution. Suppose that an initial  $\alpha_1$  is used, say  $\alpha_1 = 0.90$ . Eventually we achieve a final  $C^*$  and cannot find a feasible solution with objective less than  $0.90C^*$  i.e.,  $0.9C^*$  is a lower bound on all solutions. We either stop here, or else switch to  $\alpha_2 \geq \alpha_1$ , say  $\alpha_2 = 0.95$ . We start the search from the stored best solution corresponding to  $C^*$ , which also includes all the prohibited visitations. We are either able to find a feasible solution with objective less than  $0.95C^*$  (none less than  $0.9C^*$  exist), and repeat the process, or else conclude that none exist. We may then switch to 0.97 fathoming and eventually 100% fathom-

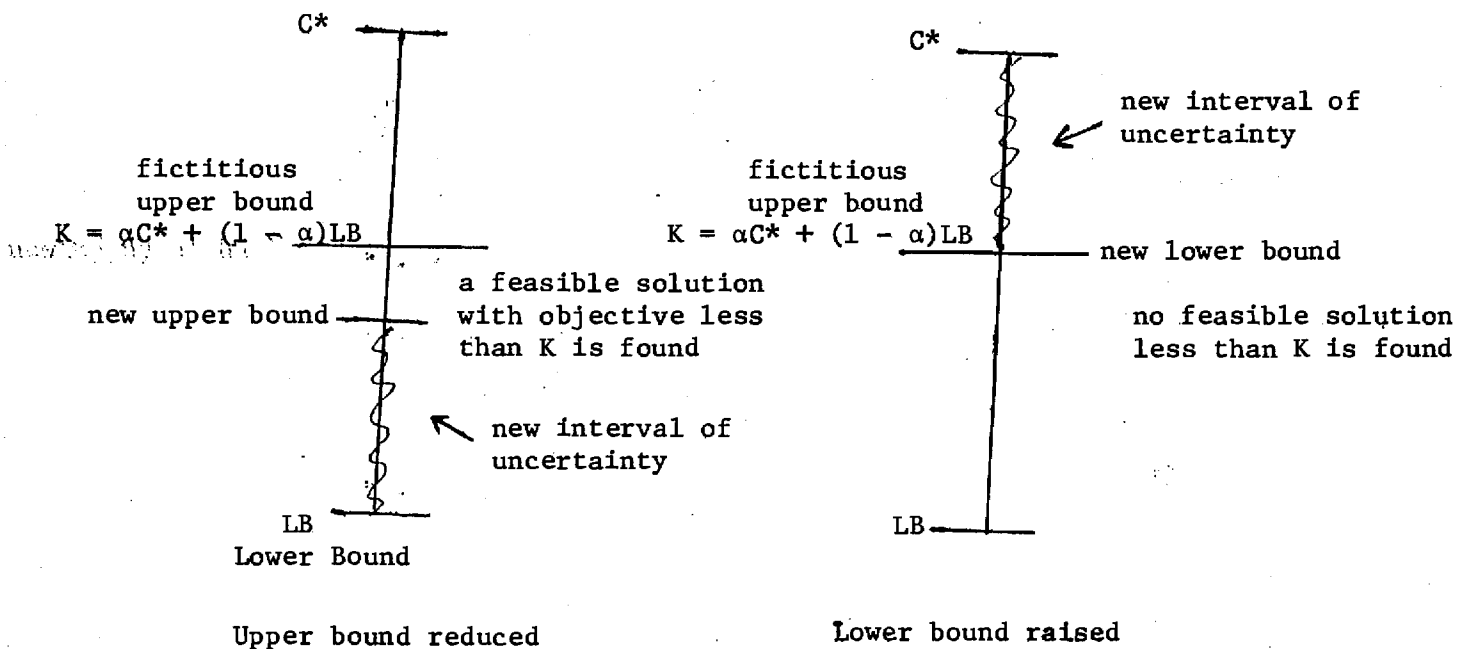
ing. This process is shown below and proved very efficient from a computational point of view.



## Method 2

This method fathoms whenever the lower bound  $LB$  is greater than or equal to  $K$ , where  $K$  is related to, but not necessarily equal to  $C^*$  ( $K \leq C^*$ ). Again we have the same interpretation as before. A partial solution is pursued if it guarantees a nontrivial improvement over what we got. We will describe a procedure to find a suitable  $K$ . At each stage of the algorithm, we have an upper bound (optimal  $\leq$  upper bound), namely  $C^*$ . A lower bound on the overall problem, namely,  $\text{Max } \theta(u,v)$  is also known. At each stage of the algorithm we have a feasible solution with objective  $C^*$ , and we know that the optimal must be less than or equal to  $C^*$ . Rather than fathoming on  $C^*$ , suppose we fathom on  $K = \alpha C^* + (1 - \alpha)LB$ , where  $\alpha \in (0,1]$ . Since  $LB < C^*$ , then  $K = \alpha C^* + (1 - \alpha)LB \leq C^*$ . Two cases are possible. In the first case, we will be able to find a tour with objective less than  $K$ . The objective of this new solution becomes the new upper bound  $C^*$  and the process is repeated. In the second case, we will not be able to find such

a solution. This automatically implies that there are no solutions with objective less than  $K$ , and hence  $K$  itself is the new lower bound. The process is repeated. From this we keep narrowing the gap between the lower and upper bounds, either by lowering the upper bound (finding an improved feasible assignment) or by raising the lower bound (by finding that there is no feasible solution with objective lower than the current lower bound). When the difference between the lower bound, and the current best objective value (upper bound) is smaller than a prescribed tolerance, we either stop, or try to find solutions with objective less than  $C^*$  itself, i.e., we switch to  $\alpha = 1$ . Of course, if the tolerance is zero, or if we switch the fathoming to  $C^*$  itself, then we will find the true optimal solution.





#### Choice of $\alpha$

$\alpha$  is any number in the interval  $(0,1]$ . Of course, if  $\alpha$  is close to 1, then we are in effect fathoming on a number very close to  $C^*$ , and the search will not speed considerably. On the other hand, if  $\alpha$  is close to zero, then we only fathom if we obtain a feasible solution very close to the overall lower bound. In this case, fathoming will be fast, but it is likely not to obtain feasible solutions less than  $K$ . The value  $\alpha = 0.5$  is recommended so that the distance of uncertainty is halved at each time. When the lower and upper bounds are close to each other, we switch to  $\alpha = 1$ .

### References

1. Bazaraa, M. S. and J. J. Goode, Lagrangian Duality In Mathematical Programming: A Unified Approach. (In preparation), Georgia Institute of Technology, Atlanta, Georgia.
2. Balas, E., An Infeasibility-Pricing Decomposition For Linear Programs, Operations Research, Vol. 14, 1966, pp. 847-873.
3. Dantzig, G. B. and P. Wolfe, The Decomposition Algorithm For Linear Programming, Econometrica, Vol. 29, No. 4, 1961.
4. Dantzig, G. B., Linear Programming and Extensions, Princeton University Press, 1963.
5. Dijkstra, E. W., A Note On Two Problems In Connexion with Graphs, Numerische Mathematik, Vol. 1, 1959, pp. 269-271.
6. Fisher, M. L., W. D. Northup, and J. F. Shapiro, Using Duality To Solve Discrete Optimization Problems: Theory and Computational Experience, Working paper OR 030-74, Operations Research Center, Massachusetts Institute of Technology, Cambridge, Massachusetts, January, 1974.
7. Geoffrion, A. M., Lagrangian Relaxation and Its Uses In Integer Programming, Working paper No. 195, Western Management Science Institute, December, 1972.
8. Geoffrion, A. M., Elements of Large Scale Mathematical Programming, Management Science, Vol. 16, No. 11, 1970, pp. 652-691.
9. Grinold, R. C., Steepest Ascent For Large Scale Linear Programs, SIAM Review, Vol. 14, 1972, pp. 447-464.
10. Held, M. and R. M. Karp, The Traveling Salesman Problem and Minimal Spanning Trees, Operations Research, Vol. 18, No. 6, 1970, pp. 1138-1162.
11. Held, M. and R. M. Karp, The Traveling Salesman Problem and Minimum Spanning Trees: Part II, Mathematical Programming, Vol. 1, 1971, pp. 6-25.
12. Held, M., P. Wolfe, and H. D. Crowder, Validation of Subgradient Optimization, IBM Report RC 4462, August, 1973.
13. Kruskal, J. B., On The Shortest Spanning Subtree Of A Graph And The Traveling Salesman Problem, Proc. Am. Math. Soc., Vol. 1, 1956, pp. 48-50.
14. Lasdon, L. S., Optimization Theory For Large Systems, The Macmillan Company, 1970.

15. Rockafellar, R. T., Convex Analysis, Princeton University Press, 1970.
16. Stoer, J. and C. Witzgall, Convexity And Optimization In Finite Dimensions I, Springer Verlag, 1970.
17. Zangwill, W. I., Nonlinear Programming: A Unified Approach, Prentice-Hall, 1969.

## Personnel

### Principal Investigator

Dr. Mokhtar S. Bazaraa  
Associate Professor  
Industrial and Systems Engineering

Organization and Coordination of the project  
Developing Quadratic Assignment and Traveling Salesman Algorithms  
Review of Programming Development

### Graduate Research Assistants

1. Virginia Ruth Khege  
Computer Program of Initial Solutions of Quadratic Assignment  
and Traveling Salesman Problems
2. Peter H. Geddes  
Applications of Quadratic Set Covering and Assignment Problems
3. Sinnan S. Tumer  
Programming of Quadratic Assignment and Set Covering Algorithms

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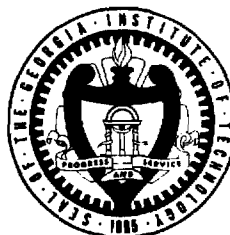
THE QUADRATIC SET COVERING (ASSIGNMENT)

PROBLEM: APPLICATIONS AND COMPUTATION

PRESENTED TO THE  
NATIONAL SCIENCE FOUNDATION  
UNDER NSF GRANT  
GK - 38337

PRINCIPAL INVESTIGATOR  
MOKHTAR S. BAZARAA

SCHOOL OF INDUSTRIAL AND SYSTEMS ENGINEERING  
GEORGIA INSTITUTE OF TECHNOLOGY  
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Principal Investigator

Mokhtar S. Bazaraa

School of Industrial and Systems Engineering

Georgia Institute of Technology

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## I. INTRODUCTION

### I.1 Summary and Highlights of the Report

Linear set covering and set partitioning problems have attracted the attention of many researchers. This is due to the simple formulation of these problems and their wide range of applications.

This research concerns itself with the quadratic set covering problem, the quadratic assignment problem, and other related problems. The significance of these problems is reflected by the flexibility of the quadratic objective, and its ability to deal with first order interaction between the variables, which leads to a large number of applications ranging from location of facilities, computerized layout design, sequencing of jobs on a machine, political districting, to selection of projects.

The following are the main results of the research.

- A unified presentation of the quadratic set covering, the quadratic assignment, and related problems. Applications of these problems in the areas of location, computerized layout, routing, project selection, sequencing, and scheduling are unified.
- Development of an algorithm for finding optimal and sub-optimal solutions of quadratic set covering problems, in the general context of layout design.
- Development of exact and heuristic algorithms for solving the quadratic assignment problem.
- Development of an exact algorithm for solving symmetric and nonsymmetric traveling salesman problems via branch and bound.

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- Development of exact and heuristic algorithms for solving the quadratic assignment problem.
- Development of an exact algorithm for solving symmetric and nonsymmetric traveling salesman problems via branch and bound.

## I.2 Organization of the Report

In Section II of the report, a classification of the quadratic set covering and related problems is presented. Different known and new applications of these problems are also presented in Section II. The remaining sections are devoted to the development of solution procedures for this class of problems. In Section III a branch and bound algorithm for solving quadratic set covering problems which arise in the context of layout design is developed. Both branch and bound and heuristic algorithms for solving quadratic assignment problems are developed in Section IV. Finally, in Section V a special procedure based on the concept of duality is developed for solving symmetric and nonsymmetric traveling salesman problems. In each case, random as well as standard problems in the literature are solved to illustrate computational effectiveness of the procedures developed. Appendices A, B, and C reproduce the algorithms of this study.

## II. THE QUADRATIC SET COVERING PROBLEM, RELATED PROBLEMS, AND APPLICATIONS

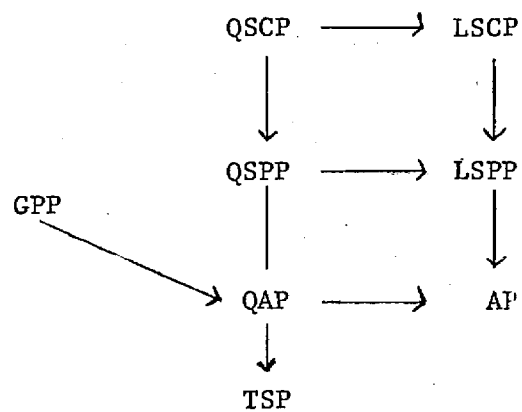
In this section we will describe the quadratic set covering problem and other related problems, and then give a number of examples illustrating the applications of these problems.

### II.1 Mathematical Descriptions

We shall consider the following problems:

quadratic set covering problem	QSCP
quadratic set partitioning problem	QSPP
linear set covering problem	LSCP
linear set partitioning problem	LSPP
generalized placement problem	GPP
quadratic assignment problem	QAP
assignment problem	AP
traveling salesman problem	TSP

All of the problems, except the GPP which is new, are well known. As the problems are precisely described below, it will be helpful to keep in mind the following diagram showing the relationships of the problems as special cases of the others in the formulations we give.



Let  $x$  be an  $n \times 1$  binary vector,  $x'$  the transpose of  $x$ ,  $A$  an  $m \times n$  matrix of zeros and ones,  $C$  an  $n \times n$  matrix, and  $e_m$  an  $m \times 1$  vector of ones. The quadratic set covering problem (QSCP) is

$$\begin{aligned} &\text{Minimize } x'Cx \\ &\text{such that } Ax \geq e_m \\ &x \text{ is a binary vector} \end{aligned}$$

and the quadratic set partitioning problem (QSPP) is

$$\begin{aligned} &\text{Minimize } x'Cx \\ &\text{such that } Ax = e_m \\ &x \text{ is a binary vector} \end{aligned}$$

If we let  $I_n$  be the  $n \times n$  identity matrix, then by choosing  $A$  to be the  $2n \times n^2$  matrix

$$A = \begin{bmatrix} e' & & & & \\ & e' & & & \\ & & \cdot & & \\ & & & \cdot & \\ & & & & e' \\ I_n & I_n & \cdot & \cdot & I_n \end{bmatrix}$$

we obtain the quadratic assignment problem (QAP) for the case of assigning  $n$  objects to  $n$  locations. We may always assume, without loss of generality, that the number of objects to be assigned is equal to the number of locations, for if this is not the case, then we may add dummy objects which make them equal but have no effect on the final solution. If the objective is of the form  $d'x + x'Cx$ , the problem is usually called a placement problem. The additional linear term represents no

generalization, however, since the quadratic objective can also handle the linear terms by placing the entries  $d_j$ 's along the diagonal of  $C$ . Therefore, we give no special considerations to placement problems as they are treated within the general class of quadratic set covering and set partitioning problems. By replacing the quadratic objective functions of the QSCP, QSPP, and QAP with a linear function, we obtain the linear set covering problem (LSCP), linear set partitioning problem (LSPP), and assignment problem (AP) respectively. The traveling salesman problem (TSP), the problem of a salesman who must visit  $n$  cities, each exactly once, and return home while minimizing the total distance of his trip will be formulated later as a special case of the QAP. The generalized placement problem (GPP) allows one to require that certain sets of objects must be assigned to certain sets of locations, or certain sets of locations must be filled by objects from certain sets of objects (Geddes [13]). As will be seen in the applications, these are very realistic restrictions, and when the GPP is formulated in terms of 0-1 variables the complications are actually helpful in obtaining the solution more efficiently. Before describing the GPP we need to introduce some notation. For a finite set  $S$  let  $|S|$  = the number of elements of  $S$  and  $P(S) = \{X | X \subset S\}$  = the power set of  $S$ . Suppose we are given two finite sets  $A$  and  $B$  with  $|A| = |B|$ . (Since  $A$  is thought of as a set of elements to be assigned to locations  $B$ , we must have  $|A| \leq |B|$ . There is no loss generality in assuming  $|A| = |B|$ , since "dummy" elements may always be added to  $A$  which do not influence the objective function.)  $\Phi = \{\psi | \psi: A \rightarrow B, \psi \text{ 1-1 map}\}$ . Assume that fixed assignment costs of elements of  $A$  to  $B$ , quantities shipped, and costs of shipping have been taken into consideration to produce a quadratic objective function, as in the QAP, and let  $c_\psi$  be the cost associated with  $\psi \in \Phi$ . Let

$R \subseteq P(A) \times P(B)$ . We say that an element  $\psi \in \Phi$  satisfies  $R$ , written  $\psi SR$ , provided:

for each  $(\hat{A}, \hat{B}) \in R$ ,  $\psi[\hat{A}] \subset \hat{B}$  if  $|\hat{A}| < |\hat{B}|$

$\psi^{-1}[\hat{B}] \subset \hat{A}$  if  $|\hat{A}| > |\hat{B}|$

$\psi[\hat{A}] = \hat{B}$  if  $|\hat{A}| = |\hat{B}|$

Here  $\psi[\hat{A}]$  denotes the image of the set  $\hat{A}$  under the map  $\psi$ . The generalized placement problem (GPP) is

$$\begin{array}{ll} \text{minimize} & c_\psi \\ \text{subject to} & \psi \in \Phi \\ \text{and} & \psi SR \end{array}$$

where  $A$ ,  $B$ ,  $R$ , and the costs  $c_\psi$  are given. Note that if  $R = \{(A, B)\}$  we simply have the QAP. Other cases, such as when  $R$  is of the form  $R = \{(\{a_{i_1}\}, \{b_{i_1}\}), \dots, (\{a_{i_s}\}, \{b_{i_s}\})\}$ ,  $s < |A|$ , will be considered in sections II.2 and II.3, where examples are given for location problems and the TSP. A procedure is also given for writing the GPP in terms of 0-1 variables.

## II.2 The Location Problems

Because of their similar mathematical structure as quadratic assignment problems, as well as analogous physical and economic interpretations, we give four applications under the general classification of location problems. These are

- a. facilities location
- b. building layout
- c. control panel layout
- d. wiring problem

Each application will be discussed in detail individually, but first the location problem in general, including properties and extensions common to all of the applications, is described. We suppose that  $n$  objects  $a_i$  ( $i = 1, 2, \dots, n$ ) are to be assigned to  $n$  locations  $b_k$

( $k = 1, 2, \dots, n$ ) in such a way that each object is assigned to one and only one location. As was mentioned earlier, it is easily seen that if we have  $m$  objects and  $n$  locations with  $m < n$ , we may supply  $n - m$  dummy objects so that there is no loss of generality in assuming the same number of objects and locations. There may or may not be a fixed cost  $f_{ik}$  associated with assigning object  $a_i$  to location  $b_k$ . There are given a fixed number of "units" to be "shipped" from object  $a_i$  to object  $a_j$ , say  $u_{ij}$ , and in general  $u_{ij} \neq u_{ji}$ . These units may be interpreted as trips made by people in the building layout problems, as we shall see, or as numbers of wire connections in the wiring problem. Shipping one unit from location  $b_k$  to  $b_\ell$  incurs a cost  $d_{k\ell}$  (usually related to distance). Define  $n^2$  variables  $x_{ik}$  ( $i = 1, 2, \dots, n$ ;  $k = 1, 2, \dots, n$ ) by

$$x_{ik} = \begin{cases} 1 & \text{if object } a_i \text{ is placed in location } b_k \\ 0 & \text{otherwise} \end{cases}$$

and let  $x' = (x_{1,1} \ x_{1,2} \ \dots \ x_{1,n} \ \dots \ x_{n,1} \ x_{n,2} \ \dots \ x_{n,n})$ .

For  $i = 1, 2, \dots, n$  and  $j = 1, 2, \dots, n$ , define  $n^2$  matrices of order  $n \times n$

$$C_{ij} = \begin{Bmatrix} C_{rs}^{ij} \end{Bmatrix} \quad (r = 1, 2, \dots, n; s = 1, 2, \dots, n)$$

$$\text{where, if } i = j, \text{ then } C_{rs}^{ij} = \begin{cases} f_{ir} & \text{if } r = s \\ 0 & \text{if } r \neq s \end{cases}$$

$$\text{if } i \neq j, \text{ then } C_{rs}^{ij} = \begin{cases} 0 & \text{if } r = s \\ u_{ij} d_{rs} & \text{if } r \neq s \end{cases}$$

$C_{rs}^{ij}$  is therefore interpreted as the cost of shipping  $u_{ij}$  units from object  $a_i$  to object  $a_j$ , if they were located in  $b_r$  and  $b_s$  respectively. Let  $C$  be the partitioned matrix



$$C = \begin{bmatrix} c_{11} & c_{12} & \dots & c_{1n} \\ c_{21} & c_{22} & \dots & c_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1} & c_{n2} & \dots & c_{nn} \end{bmatrix}$$

Let  $e_n$  be a  $n \times 1$  vector of ones and  $I_n$  be the  $n \times n$  identity matrix.

Let  $H$  be the  $2n \times n^2$  assignment matrix

$$H = \begin{bmatrix} e_n' & & & \\ & e_n' & & \\ & & \ddots & \\ & & & e_n' \\ I_n & I_n & \dots & I_n \end{bmatrix}$$

The location problem (quadratic assignment problem) is to

$$\text{minimize } x' C x$$

$$\text{subject to } Hx = e_{2n}$$

$$x \text{ binary}$$

It should be noted that a large number of elements of the  $C$  matrix were set equal to zero since they correspond to assignments not permitted by  $Hx = e_{2n}$ . The first  $n$  rows of equations represented by  $Hx = e_{2n}$  require that each object is assigned to one location, and the remaining  $n$  equations require that each location is assigned to one object; that is to say we have a one-to-one correspondence between objects and locations.

Now, for example, the object  $a_1$  cannot be assigned to locations  $b_1$  and  $b_3$  simultaneously, so the first row third column entry of  $C_{11}$ ,  $C_{13}^{11}$ , was set equal to zero. Actually, any number could be placed

in this entry, since it would never contribute to the value of the objective function  $x'Cx$  for any feasible  $x$ .

There are two extensions of the location problem which fall into the case of the GPP. In one of these we may require a few selected objects to be assigned to particular locations. For example, we may be given a collection of factories and warehouses already built and wish to add some new ones. A slightly different problem occurs when certain sets of objects must be assigned to certain sets of locations, or certain sets of locations must be filled by objects from given sets. In the second instance, we still have some freedom to locate the object. An example is that the tuning knob on a radio has to be, say, on the outside in front and not buried in the middle of the set, but there are several locations available. As another example, consider the problem of assigning ten plants of various sizes, some of which require a water supply, to sites of different sizes, some of which have water supplies. This type of constraint may also be used to keep certain objects either close together or far apart. In a warehouse certain chemicals cannot be stored together for safety reasons such as fire and explosion. Certain areas in a warehouse may be unsuitable for storing some materials because of temperature or humidity, while some materials may be stored in any area of the warehouse. We may wish to require certain rooms in an office building to be located on an outside wall. In the design of electrical equipment some components may fail to function properly if located too far apart, while others fail when located too close together. An alternative to using the GPP to handle these complications would be to add constraints which simply require the distance between the locations containing objects be greater than (or less than) some fixed number.

a. Facilities Location

We begin by describing a simple example in which three plants  $a_i$  ( $i = 1, 2, 3$ ) are to be assigned to three locations  $b_k$  ( $k = 1, 2, 3$ ). The fixed costs  $f_{ik}$  of assigning plant  $a_i$  to location  $b_k$  are given in Table II.1. Table II.2 gives the number of units  $u_{ij}$  to be shipped from plant  $a_i$  to plant  $a_j$ , and Table II.3 gives the unit costs  $d_{kl}$  of shipping from location  $b_k$  to  $b_l$ .

		Location $b_k$		
		1	2	3
Plant $a_i$	1	200	300	250
	2	400	350	300
	3	600	700	650

Table II.1 Values of  $f_{ik}$

		Plant $a_j$		
		1	2	3
Plant $a_i$	1	--	50	100
	2	60	--	80
	3	40	70	--

Table II.2, Values of  $u_{ij}$

		Location $b_l$		
		1	2	3
Location $b_k$	1	--	2	3
	2	1	--	2
	3	3	3	--

Table II.3 Values of  $d_{kl}$

We compute the C and H matrices described above:

$$C = \begin{bmatrix} c_{11} & c_{12} & c_{13} \\ c_{21} & c_{22} & c_{23} \\ c_{31} & c_{32} & c_{33} \end{bmatrix}$$

200	-	-	-	100	150	-	200	300
-	300	-	50	-	100	100	-	200
-	-	250	150	150	-	300	300	-
-	120	180	400	-	-	-	160	240
60	-	120	-	350	-	80	-	160
180	180	-	-	-	300	240	240	-
-	50	120	-	140	210	600	-	-
40	-	80	70	-	140	-	700	-
120	120	-	210	210	-	-	-	650

$$H = \begin{bmatrix} 1 & 1 & 1 & & & & & & \\ & & & 1 & 1 & 1 & & & \\ & & & & & & 1 & 1 & 1 \\ 1 & & & 1 & & & 1 & & \\ & 1 & & & 1 & & & 1 & \\ & & 1 & & & 1 & & & \\ & & & 1 & & & & & 1 \end{bmatrix}$$

As before, for  $i = 1, 2, 3$ ;  $k = 1, 2, 3$ ; let

$$x_{ik} = \begin{cases} 1 & \text{if plant } a_i \text{ is assigned to location } b_k \\ 0 & \text{otherwise} \end{cases}$$

$$x' = (x_{11} \ x_{12} \ x_{13}, \ x_{21} \ x_{22} \ x_{23}, \ x_{31} \ x_{32} \ x_{33}), \text{ and}$$

$$e'_6 = (1 \ 1 \ 1 \ 1 \ 1 \ 1).$$

The location problem is then

$$\text{minimize } x'Cx$$

$$\text{subject to } Hx = e_6$$

and  $x$  is a binary vector

It should be emphasized that in this problem the quantities shipped between plants,  $u_{ij}$ , were given and fixed, which in more general instances is not always the case. There are a number of complications which may be introduced. For example, the shipment of goods between plants, if placed in certain locations, may require the construction of a road or pipeline, and the fixed cost of such construction must be taken into consideration. It is also possible that more than one plant may be placed in one location. A very common problem is that we already have a number of plants built and wish to add a few more. Suppose we are considering  $n$  plants  $a_i$  and  $n$  locations  $b_k$ , but  $s < n$  of these plants are already built. In particular, assume that for  $j = 1, 2, \dots, s$  we have plant  $a_{i_j}$  built in location  $b_{i_j}$ . We can formulate this problem as a GPP by letting

$$R = \{(\{a_{i_1}\}, \{b_{i_1}\}), (\{a_{i_2}\}, \{b_{i_2}\}), \dots, (\{a_{i_s}\}, \{b_{i_s}\})\}$$

A second more complicated example of the GPP is given, and the relationship of the 0-1 variables shown. Let  $A = \{a_1, a_2, \dots, a_{10}\}$  be a set of ten plants and  $B = \{b_1, b_2, \dots, b_{10}\}$  be a set of ten locations. We are given that  $u_{ij}$  units are to be shipped from plant  $a_i$  to plant  $a_j$ , the cost of shipping one unit from location  $b_k$  to  $b_l$  is  $d_{kl}$ , and the cost of placing plant  $a_i$  in location  $b_k$  is  $f_{ik}$ .

$$\text{Let } A_1 = \{a_8, a_9, a_{10}\}$$

$$B_1 = \{b_8, b_9, b_{10}\}$$

$$A_2 = \{a_4, a_5, a_6, a_7, a_8\}$$

$$B_2 = \{b_1, b_2, b_3, b_4, b_5, b_8, b_9\}$$

$$A_3 = \{a_1, a_2, a_3, a_4, a_5, a_9\}, \text{ and}$$

$$B_3 = \{b_2, b_7, b_8\}$$

Let  $R = \{(A_1, B_1), (A_2, B_2), (A_3, B_3)\} \subset P(A) \times P(B)$

We define  $C$  and  $H$  as before and obtain the problem:

$$\text{minimize } x'Cx \quad \text{II.1}$$

$$\text{subject to } Hx = e_{20} \quad \text{II.2}$$

$$x \text{ binary} \quad \text{II.3}$$

and elements from  $A_1$  must be placed in locations of  $B_1$ , and II.4a

only those locations

elements from  $A_2$  must be placed in locations of  $B_2$  II.4b

elements placed in  $B_3$  locations must come from  $A_3$  II.4c

Statements II.1, II.2, II.3 make up a QAP, and II.4abc tell us that

$$x_{ij} = 0 \text{ for } i \text{ and } j \text{ such that } \left\{ \begin{array}{ll} a_i \in A_1 \text{ and } b_j \in B \setminus B_1 & \text{II.4a} \\ \text{or } a_i \in A \setminus A_1 \text{ and } b_j \in B_1 & \\ a_i \in A_2 \text{ and } b_j \in B \setminus B_2 & \text{II.4b} \\ a_i \in A \setminus A_2 \text{ and } b_j \in B_2 & \\ a_i \in A_3 \text{ and } b_j \in B \setminus B_3 & \text{II.4c} \\ a_i \in A \setminus A_3 \text{ and } b_j \in B_3 & \end{array} \right.$$

(Here  $A \setminus A_1$  is the complement of  $A_1$  in  $A$ , etc.)

Interpretations of these constraints might be: plants of set  $A_2$  require a water supply which is available only in locations of set  $B_2$ ; locations of  $B_3$  are small and must be occupied by plants not requiring much space, those of  $A_3$ ; and plants of  $A_1$  must be built together in a certain region  $B_1$ .

## b. Building Layout

The building layout problem is quite similar to the facilities location problem except that we are concerned with what goes on between various rooms in a building rather than shipments between plants. One objective may be to minimize the time employees spend walking between offices or rooms by finding optimal locations for the offices. For

example, in the case of a hospital, we have a wide collection of employees such as surgeons, nurses, student nurses, etc., who collect various salaries and find it necessary (according to their job) to travel with varying degrees of frequency between different rooms such as an emergency room, operating room, scrub-up room, or sterilizing room. Now it has been pointed out that, on the average, people in a typical hospital spend 38 percent of their time walking between rooms, and taking relative salaries into account, this represents 34 percent of the cost for staff salaries. If we assume that careful planning can reduce this figure by one quarter, a savings equivalent to 8 1/2 percent of the cost for staff salaries is realized. This may amount to a substantial sum of money, especially over a period of many years (Whitehead and Elders [30]).

There are, however, some special problems in building layout which do not occur in any analogous manner in facilities location, control panel layout, or the wiring problem. We begin in a typical building layout problem by making a list of the rooms, and the frequency with which employees of different salaries travel between the rooms. For each pair of rooms  $a_i$  and  $a_j$  we may then compute a cost factor  $u_{ij}$  equal to the sum of the products of the number of people traveling between the two rooms and their salaries (say, in dollars/hours). The movements are then examined to see if they can be eliminated as unnecessary or new facilities or organization could lead to their elimination. For example, a recovery ward may be added so the surgeon or anaesthetist doesn't need to visit patients in their own room right after an operation. A set of location units  $b_k$  is determined. Since the rooms may vary in size, some may require only one location while others require several (adjacent locations). Instead of distance  $d_{kl}$  between locations  $b_k$  and  $b_l$ , we let  $d_{kl}$  represent

the time (say, in fractions of hours) it takes to travel from  $b_k$  to  $b_l$ . Of course, a difficulty arises when a room occupies several locations, and in this case we may arbitrarily take the shortest (or longest) distance between the locations used by the rooms.

Once a placement has been determined, refinements may be needed such as the addition of a corridor to help maintain aseptic conditions in the operating room, or even changes for aesthetic purposes. It should be pointed out then, that the solution of the building layout problem as a QAP (or GPP) yields only an approximate solution to the real problem. Some factors such as the cost of air ducts, plumbing, and wiring between rooms we have neglected, because in most instances the differences are small for various layouts.

#### c. Control Panel Layout

One of the objectives in designing an instrument or control panel may be to minimize eye travel distance. Assume we are assigning  $n$  instruments  $a_i$  to  $n$  locations  $b_k$ . Using notation similar to before, we let  $u_{ij}$  be the frequency of transitions from instruments  $a_i$  to  $a_j$  and  $d_{kl}$  the Euclidean distance between locations  $b_k$  and  $b_l$ . As before, the cost and assignment matrices  $C$  and  $H$  may be computed to obtain a QAP:

$$\begin{aligned} &\text{minimize} && x' C x \\ &\text{subject to} && H x = e_{2n} \\ &&& x \text{ binary} \end{aligned}$$

This problem is discussed in (Dorris [8, p. 18-21]), while a linear case is given in (Freund and Sadosky [9]).

#### d. The Wiring Problem

Beginning in the late 1950's, a problem associated with the construction of computers, now known as the backboard wiring problem, began to



receive attention (Loberman and Weinberger [20], and Steinberg [25]).

The problem is to find a placement (assignment) for a number of electrical components, which must be interconnected, on a board so that the total length of wire used to make the connections is minimal. An assumption is made that minimal total wire length is the best approximation for satisfying a number of goals for good circuit design, which include:

- reducing capacitance
- reducing delay line effects
- distribution of heat sinks
- avoidance of excessive wire build-up on routing channels
- avoiding cross-talk without excessive use of expensive shielding
- neatness to reduce construction and maintenance costs
- using smallest amount of wire

We suppose a set of  $n$  electrical components  $a_i$  are given which must be assigned to a set of  $n$  locations  $b_k$ . As before, without loss of generality, it is assumed the number of components equals the number of locations, and they will be assigned in a one-to-one manner. Let  $u_{ij}$  be the number of wires connecting components  $a_i$  and  $a_j$ . The distance  $d_{kl}$  between  $b_k$  and  $b_l$  will, of course, depend on the wiring channels used and may be either Euclidean or "street" distance. The cost and assignment matrices may be found, as before, to give the problem:

$$\begin{aligned} &\text{minimize} && x'Cx \\ &\text{subject to} && Hx = e_{2n} \\ &&& x \text{ binary} \end{aligned}$$

Example: The best known example, and, in fact, about the only real problem which appears to have been studied in detail by many authors, is the one given in (Steinberg [25], and (Graves and Whinston [15])). It is pointed out in (Hanan and Kurtzberg [10]) that there is a need for studying additional realistic problems. Here we give a small example just to illustrate the ideas involved, the structure of the matrices, and some

$\mathbb{K}^2$ 

i \ j	1	2	3	4
1	0	3	7	1
2	3	0	2	5
3	7	2	0	3
4	1	5	3	0

Table II.4 Values of  $u_{ij}$

k \ 1	1	2	3	4
1	0	1	1	$\sqrt{2}$
2	1	0	$\sqrt{2}$	1
3	1	$\sqrt{2}$	0	1
4	$\sqrt{2}$	1	1	0

Table II.5 Values of  $d_{kl}$   
with Euclidean Distance

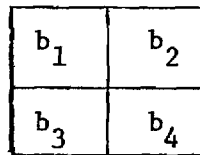


Figure II.1

The assignment matrix  $H$  is

[illegible]

The cost matrix  $C$  could be computed as before, but since the  $(u_{ij})$  and  $(d_{kl})$  matrices are symmetric, we may write  $C$  as the partitioned matrix

$$C = \begin{pmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{pmatrix}$$

$$\text{where } c_{ij} = \begin{pmatrix} c_{ij} \\ c_{rs} \end{pmatrix} \quad \begin{matrix} r = 1, 2, 3, 4 \\ s = 1, 2, 3, 4 \end{matrix}$$

$$\text{and } c_{rs}^{ij} = \begin{cases} u_{ij} d_{rs} & \text{if } i < j \text{ } (i = 1, 2, 3, j = 2, 3, 4) \text{ and } r \neq s \\ 0 & \text{otherwise} \end{cases}$$

So C is the 16 x 16 matrix whose non-zero elements are shown:

$$C = \begin{bmatrix} \begin{matrix} 0 & 3 & 3 & 3\sqrt{2} \\ 3 & 0 & 3\sqrt{2} & 3 \\ 3 & 3\sqrt{2} & 0 & 3 \\ 3\sqrt{2} & 3 & 3 & 0 \end{matrix} & \begin{matrix} 0 & 7 & 7 & 7\sqrt{2} \\ 7 & 0 & 7\sqrt{2} & 7 \\ 7 & 7\sqrt{2} & 0 & 7 \\ 7\sqrt{2} & 7 & 7 & 0 \end{matrix} & \begin{matrix} 0 & 1 & 1 & \sqrt{2} \\ 1 & 0 & \sqrt{2} & 1 \\ 1 & \sqrt{2} & 0 & 1 \\ \sqrt{2} & 1 & 1 & 0 \end{matrix} \\ \begin{matrix} 0 & 2 & 2 & 2\sqrt{2} \\ 2 & 0 & 2\sqrt{2} & 2 \\ 2 & 2\sqrt{2} & 0 & 2 \\ 2\sqrt{2} & 2 & 2 & 0 \end{matrix} & \begin{matrix} 0 & 5 & 5 & 5\sqrt{2} \\ 5 & 0 & 5\sqrt{2} & 5 \\ 5 & 5\sqrt{2} & 0 & 5 \\ 5\sqrt{2} & 5 & 5 & 0 \end{matrix} & \begin{matrix} 0 & 3 & 3 & 3\sqrt{2} \\ 3 & 0 & 3\sqrt{2} & 3 \\ 3 & 3\sqrt{2} & 0 & 3 \\ 3\sqrt{2} & 3 & 3 & 0 \end{matrix} \end{bmatrix}$$

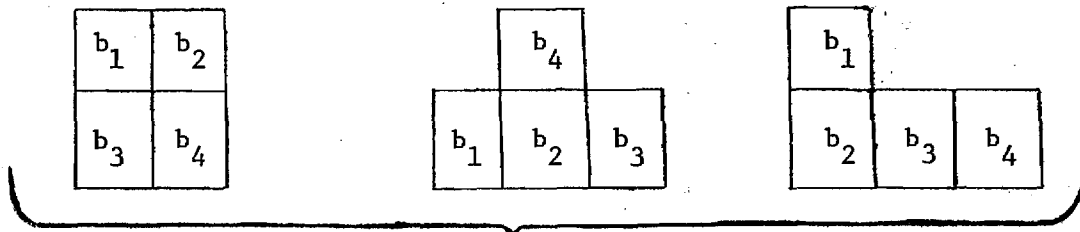
The problem is

$$\text{minimize } x' C x$$

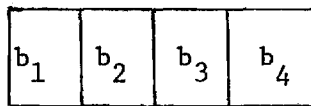
$$\text{subject to } Hx = e_8$$

$$x \text{ binary}$$

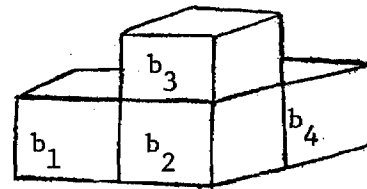
Some Extensions The above problem may be altered by introducing weights to the  $u_{ij}$ 's, as was done in the hospital example of the building layout problem, or by changing the  $(d_{kl})$  matrix. We may wish to change the  $(d_{kl})$  matrix to correspond to such physical arrangements of locations as



2 dimensional arrangements



1 dimensional



or possibly a 3 dimensional layout

If  $\mathcal{D}$  is a set of matrices

$$\mathcal{D} = \{D = (d_{kl})\}$$

we may wish to consider the problem of finding the best arrangement of locations as well:

$$\begin{array}{lll} \text{minimize} & \text{minimize} & x' C_D x \\ D \in \mathcal{D} & Hx = e & \\ & x \text{ binary} & \end{array}$$

(Here  $C_D$  indicates the cost matrix depends on  $D$ ).

As was done in the facilities location problem, we may use the GPP to require certain sets of components be located in specified sets of locations. This may be useful to keep the components of one section, say an amplification or power section, of a piece of electrical equipment together, as an aid to construction or maintenance.

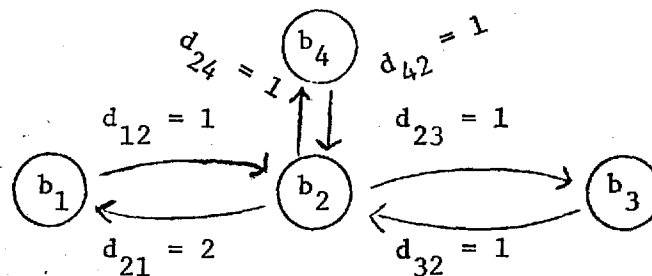
### II.3 The Traveling Salesman Problem

Suppose a salesman is given a list of  $n$  cities connected by a number of roads of known distance. His problem, known as the traveling salesman problem, is to find a route of minimal length he can follow so as to visit each city once and only once, and return to where he started (Hu [17, pp. 270-271] and Garfinkel and Nemhauser [11, pp. 354-361]).

#### a. The Classical Traveling Salesman Problem

There are several different versions and formulations of the TSP. We are concerned with starting the problem as a QAP, but first mention, for clarification, some of the variations. First, we are considering what is called the closed tour version. This simply means that the salesman must end up where he started. An alternative, is the open tour version in which the salesman starts in a given city, visits each city once and only once, but does not return to the original city. It is easy to convert the closed tour version to the open tour, by dividing the starting city into two cities. We are just going to consider the closed tour here.

Now the following example is a simple TSP which has no solution. We are given four cities  $b_k$  ( $k = 1, 2, 3, 4$ ) and some roads of known distances  $d_{kl}$  between them as shown below.



It is easy to see that it is impossible to visit all four cities and not visit  $b_2$  three times. There is a modification of the TSP described in b. below which allows for the possibility of passing through cities, but as the classical TSP, this problem has no solutions. We also note that  $d_{12} = 1$  and  $d_{21} = 2$ , so that  $d_{12} \neq d_{21}$ . It is not necessary that the distances between two cities be the same each way. The idea of a subtour comes from formulations of the TSP in which variables  $x_{ij}$  are used:

$$x_{ij} = \begin{cases} 1 & \text{if the salesman travels from } b_i \text{ to } b_j \\ 0 & \text{otherwise} \end{cases}$$

Unless constraints are added to prevent it, a "solution" may occur consisting of more than one trip, as shown below:



The trips 1-2-3-1 and 4-5-6-7-4 are called subtours, and obviously cannot be a solution to any problem in which a salesman must begin and end his trip in the same city. A tour is a sequence for visiting each city exactly once and ending up in the starting city. In the formulation we now give for the TSP as a QAP, the problem of subtours does not arise.

Suppose we are given  $n$  cities  $b_k$  ( $k = 1, 2, \dots, n$ ) with distances  $d_{k\ell}$  from city  $b_k$  to  $b_\ell$ . We assume  $d_{k\ell} \geq 0$ , and in cases where there is no  $d_{k\ell}$  given (as when there is no direct road from  $b_k$  to  $b_\ell$ ) we assign some large number  $M$ . We think of the TSP as a facilities location problem in which  $n$  plants  $a_i$  ( $i = 1, 2, \dots, n$ ) are to be assigned to the  $n$  cities  $b_k$  ( $k = 1, 2, \dots, n$ ). The units shipped from  $a_i$  to  $a_j$ ,  $u_{ij}$  are given by:

- i) for  $i = 1, 2, \dots, n-1$   $u_{i \ i+1} = 1$
- ii) for  $i = n$   $u_{n \ 1} = 1$
- iii) all others  $u_{ij} = 0$

and we wish to minimize the total cost. Constructing cost and assignment matrices  $C$  and  $H$  as before, we have the TSP stated as the QAP

$$\begin{aligned} &\text{minimize} && x' C x \\ &\text{subject to} && H x = e_{2n} \\ &&& x \text{ binary} \end{aligned}$$

A variable  $x_{ik} = 1$  in the final solution  $x$  means city  $k$  is visited as the  $i$ th city.

#### b. Modifications of the Traveling Salesman Problem

As noted in a. above there are some TSP's which have no solution because it is impossible to visit each city exactly once. There are other examples in which we could reduce the distance traveled if we could "pass through" a city, i. e., visit it more than once. Such an example is shown below in Figure II.2.

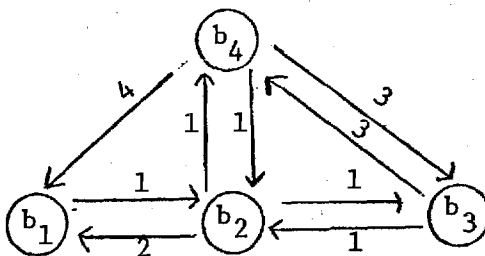


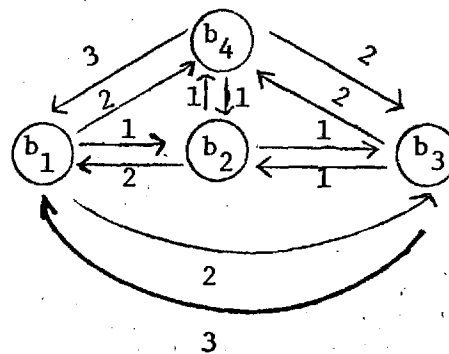
Figure II.2

The shortest route to visit all cities by the TSP is 1-2-3-4-1 and has length 9. If we are permitted to visit  $b_2$  three times, we could choose the route 1-2-4-2-3-2-1 which has length 7. One method to avoid having

the larger solution to the TSP forced upon us, which would be desirable in most real applications, is to construct an expanded TSP as follows. Given  $n$  cities and some values  $d_{k\ell}$  representing the distances of roads from city  $b_k$  to  $b_\ell$ , define values of  $\tilde{d}_{k\ell}$  for all  $k = 1, 2, \dots, n$ ;  $\ell = 1, 2, \dots, n$  by

$$\tilde{d}_{k\ell} = \begin{cases} \text{the shortest distance from} \\ \text{city } b_k \text{ to } b_\ell & \text{if } k \neq \ell \\ \infty & \text{if } k = \ell \end{cases}$$

(We will need to keep track of the shortest paths when stating the solution of the original problem from the expanded problem). The example in Figure II.2 would then become



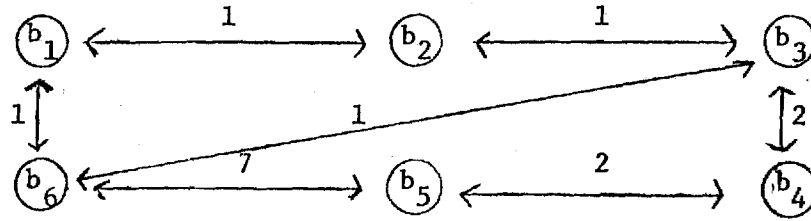
with the optimal solution 1-2-4-3-1 and distance 7. If the original problem in Figure II.2 had satisfied the triangle inequality for distances\*

$$d_{ij} + d_{jk} \geq d_{ik}$$

and we had been given values for all  $d_{k\ell}$  ( $k = 1, 2, \dots, n$ ;  $\ell = 1, 2, \dots, n$ ), then the optimal routing wouldn't require visiting  $b_2$  three times and the expanded TSP wouldn't have been needed (Hu [17, pp. 270-3]). The following example is one in which the given  $d_{k\ell}$  satisfies the triangle inequality, but a city is visited twice to obtain the shortest route 1-2-3-4-5-4-3-6-1 (length 12). The reason here is that not all the inter-distances among the cities are given.

\*It doesn't:  $2 = 1 + 1 = d_{3,2} + d_{2,4} < d_{3,4} = 3$





We now mention some extensions of the TSP. In addition to being given  $n$  cities  $b_k$  ( $k = 1, 2, \dots, n$ ) and some distances  $d_{k\ell}$  between cities  $b_k$  and  $b_\ell$ , we may be given times  $t_{k\ell}$  between cities  $b_k$  and  $b_\ell$  and due dates  $\tau_k$  by which a city  $b_k$  must be visited. For a given tour  $\psi$ , let  $b_{\psi(i)}$  be the  $i$ th city visited, and suppose  $b_{\psi(i)}$  is visited at time

$$\hat{\tau}_{\psi(i)} = t_{\psi(0)\psi(1)} + t_{\psi(1)\psi(2)} + \dots + t_{\psi(i-1)\psi(i)}$$

for  $i = 1, 2, \dots, n$ . If we define the tardiness in visiting a city as  $\hat{\tau}_k - \tau_k$ , one objective may be to find a tour  $\psi$  which minimizes the maximum tardiness. Another problem may be to minimize a linear combination of the distance traveled and the maximum tardiness. The GPP may be used to make such requirements as the first city visited must be one of three selected cities, or the last two cities visited must include a certain given city.

In another modification of the TSP, considered by (Srivastava et. al., [24]), the set of cities  $B = \{b_1, b_2, \dots, b_n\}$  is divided into a home city and a family of  $s$  collectively exhaustive and mutually exclusive sets of cities:  $B = \{\text{home city}\} \cup \bigcup_{i=1}^s B_i$ . The problem is to determine an optimal tour, starting at the home city, which includes at least one city from each of the sets  $B_i$  ( $i = 1, 2, \dots, s$ ). An application of this modification, in which a salesman has to complete multiple jobs subject to competition, is given in (Garg, et. al., [12]).

c. Production Scheduling Application

Suppose we have  $n$  tasks  $b_k$  ( $k = 1, 2, \dots, n$ ) which are to be performed on some machine. A change over time  $d_{k\ell}$  is required to change the machine from doing task  $b_k$  to task  $b_\ell$ . The problem of selecting a sequence in which to perform the tasks and return the machine to its original state so the sum of changeover times is minimized is a TSP (Maxwell [21]).

An  $n$ -job,  $m$ -machine scheduling problem is modelled as a TSP in (Wisner [31]).

d. The Multi-Traveling Salesmen Problem, Its Generalizations, and Some Applications

The TSP is actually a special case of the multi-traveling salesmen problem (M-TSP) (Svestka and Huckfeldt [26]). In the M-TSP we are given  $n$  cities  $b_k$  ( $k = 1, 2, \dots, n$ ), one of which is designated as a "home city", and distances  $d_{k\ell}$  between cities  $b_k$  and  $b_\ell$ . There are  $m$  salesmen instead of one, and the problem is to determine routes for each of the  $m$  salesmen to follow so that every city (except the home city) is visited exactly once by exactly one salesman, the home city is included in each salesman's route, and the total distance traveled by all salesmen is a minimum. We note that in the M-TSP a "home city" is designated beforehand. Three alternatives are to designate several home cities, as is done in the multi-terminal delivery problem (Tillman and Hering [27]), or allow for either one or several undesignated home cities which will be determined as part of the problem. (Not requiring a home city actually falls into the last two cases). Another generalization of the M-TSP is given in (Zak [32]) where each salesman may have a different distance matrix.

We now give some applications of the M-TSP and its generalizations.

### The Bank Messenger Scheduling Problem

In this problem a crew of messengers collects deposits at branch banks and returns them to a central office for processing. If there are  $m$  messengers and  $n$  branches, this is the M-TSP. The Cleveland Trust Company, Cleveland, Ohio, has actually made use of the M-TSP to determine routes for its messengers (Svestka and Huckfeldt [26]).

### Printing Press Scheduling for Multi-Edition Periodicals

A printing press scheduling problem is given as a 3-traveling salesmen problem with a number of side constraints in (Gorenstein [14]). But even a simplified version of this problem is not solvable for problem sizes with practical applications.

### Scheduling Excavators Movements

The scheduling of the movements of excavators in mine pits is given in (Tsoi, et. al., [28]) as a M-TSP. Cases with route length and cycle constraints are given.

### Single Terminal Delivery Problem

In this problem delivery trucks supply customers from a single terminal. This is a M-TSP, with the restriction that the trucks have limited capacity and need to return to the home city (terminal) to refill.

### Multiterminal Delivery Problem

This is an application of the generalization of the M-TSP in which several home cities are designated. While in the single terminal delivery problem we assign customers to routes so that the total distance is minimized, in the multiterminal delivery problem we need to assign customers to routes and routes to terminals. This problem and references for the single terminal case may be found in (Tillman and Hering [27]).

## II.4 Applications of the QSCP and Related Problems

In this section we give a number of examples of set covering problems, most of which were originally formulated as LSCP (or LSPP). These include the delivery problem, location of fire hydrants, airline crew scheduling, assembly line balancing problem, information retrieval, project selection, capital budgeting, and attack/defense networks. A general structure common to all of these applications is first given, then each application is considered separately. The project selection problem is considered in detail and some directions are indicated for the formulation of a more realistic model.

We recall that the QSCP was

$$\begin{array}{ll}\text{minimize} & x'Cx \\ \text{such that} & Ax \geq e_m \\ & x \text{ is a binary vector}\end{array}$$

where  $x$ ,  $C$  and  $A$  were described in II.1.  $A$  was an  $m \times n$  matrix. In the applications each of the  $m$  rows is identified with either a client (delivery problem), a flight segment (airline crew scheduling), a task (assembly line balancing), or some activity. Various subsets of the clients flight segments, tasks, or activities are selected to form (respectively) routes, trips, assembly stations, or projects, which are identified with the columns of  $A$ . For example, in the delivery problem, if client  $i$  is included in route  $j$ , then  $a_{ij} = 1$ , if not  $a_{ij} = 0$ .  $e_m$  is a vector of ones indicating that all clients must be included in the final solution. In the set covering problem, a client may be served more than once. If we want to restrict a client to one service in a problem, the partitioning problem would be used. In the LSCP and LSPP, costs are assigned to the routes, trips, assembly stations, or projects (i. e., the columns of  $A$ ).

The quadratic versions of these problems permit additional costs or discounts for various pairs of columns selected. In some applications there are good reasons why two activities should cost less than the sum of the costs of the activities individually, while in other cases there appears to be no reason for extending the linear formulation to a quadratic one.

a. Routing Problems: Truck Delivery and School Bussing

In the truck delivery and school bussing problems we are concerned with the delivery or pick-up of items, or, in the case of school bussing, children. There are a number of delivery or pick-up stations, each of which must be visited at least once, and a central terminal or school from which the items are sent out or collected. Various subsets of the delivery stations are formed to make up routes, and each route has a cost associated with it. The problem of selecting a set of routes so that each station is included in at least one route is a LSCP, or, if reasons can be found for quadratic interaction of the costs, a QSCP. In the case of school bussing one reason to use a QSCP might be to help achieve a racial balance. The question remains of how to do this. This would allow certain combinations of routes to be desirable and others to be undesirable. Since trucks or buses have limited capacity, the number of items picked up or delivered on a route is limited; also, since the total number of available trucks is limited (not allowing multiple trips for a vehicle), we may need a solution in which the total number of routes selected is less than or equal to the total number of vehicles (Balinski and Quandt [33] and Murty [22]).

b. Location of Fire Hydrants

The problem of locating fire hydrants is discussed in (Murty [22]). We consider a network whose arcs are streets, and nodes are intersections

of the streets. The problem is to select a subset of the nodes for placement of fire hydrants so each street has at least one fire hydrant. This problem is probably best left as a LSCP.

c. Airline Crew Scheduling

In the airline crew scheduling problem flight segments are selected to form trips, and the idea is to select trips to cover all flight segments. This is a LSCP. See (Murty [22] and Arabeyre, et. al., [2]).

d. Assembly Line Balancing

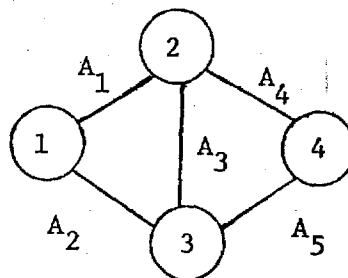
Tasks are combined to be performed at assembly stations in the assembly line balancing problem. The problem is to determine which assembly stations are to be set up. This is a partitioning problem, since a task is to be performed exactly once. Complications arise in the form of precedence relations on tasks and due dates by which tasks must be performed. (Murty, [22] and Salvesan [23]).

e. Information Retrieval

The information retrieval problem is how to select a number of tapes from a collection of tapes, each of which contains subsets of required information, so that all desired information may be found by searching a minimal length of tape. This is a LSCP and there appears to be no advantages in considering it as a QSCP. See (Day [7]).

f. Attack/Defense Networks

Attack and defense of networks may be formulated as QSCP. Consider the network shown below



The nodes in the network are thought of as strategic points; while the arcs represent communication lines such as roads, bridges, pipelines, etc. The objective is to attack the network in a optimal manner so that nodes 1 and 4 are separated. If  $A_i$  represents arc  $i$  and  $x_i$  the associated variable, then  $x_i = 1$  means  $A_i$  is destroyed and  $x_i = 0$  means  $A_i$  is not destroyed. The constraint matrix which requires every path from node 1 to node 4 be cut at least once is

$$\begin{array}{ccccc}
 A_1 & A_2 & A_3 & A_4 & A_5 \\
 \left[ \begin{array}{ccccc}
 1 & 0 & 0 & 1 & 0 \\
 0 & 1 & 0 & 0 & 1 \\
 1 & 0 & 1 & 0 & 1 \\
 0 & 1 & 1 & 1 & 0
 \end{array} \right] & \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} & \geq & \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}
 \end{array}$$

If a cost is assigned to destroying each arc, a linear objective function may be obtained to give a LSCP. However, the cost of destroying two arcs, say  $A_1$  and  $A_5$ , may be greater than the sum of the costs of destroying them individually. For example, they may be far apart and require some scarce resource to destroy. In this case a quadratic objective function  $x'Cx$  is used to consider such interactions. The element  $c_{15}$  would be positive and represent the additonal cost of cutting  $A_1$  and  $A_5$ . The attack problem is a good example of a LSCP which is better modelled as a QSCP. The defense problem, how to optimally defend a network which will be attacked, by the construction of more arcs subject cut limitations, is also a QSCP. (Bellmore and Ratliff [4,5] and Jarvis [18]).

#### g. Project Selection and Capital Budgeting

We consider the problem where a number of projects are combined to form activities. A set of activities is to be selected in an optimal

manner so that all required projects are included among the selected activities. Since it would probably be desired that a project is not included more than once, a partitioning problem would be used. Since there may be advantages of selecting certain pairs of activities, a QSCP is used.

To give some insight into the discount cost arising from the selection of two activities, consider the interaction which occurs in construction activities. There are three primary reasons why two activities are sometimes cheaper than the sum of the individual costs:

1. a materials discount When two activities contain projects which use the same type of material, a savings may be realized by buying this material in bulk. There is a limit to this savings, e. g., three \$5,000 projects may lead to a savings, but the addition of two more \$5,000 projects may not be significant.
2. already having equipment on the construction site To move a piece of machinery may cost \$400, \$100 to get a truck on the site and \$300 to move the machine. Perhaps three machines could be moved at the same time for  $100 + 3(300) = \$1000$  or a \$200 savings over  $3(400) = \$1200$ .
3. insurance savings (four jobs cost only a little more than one).

Unfortunately, there are also three major reasons why the QSPP doesn't apply in many cases. First, we assume we know all of the projects (and activities) from the start, while in real problems new projects arise as work proceeds on old ones. Then we assume all the projects are funded ahead of time and from one source. In practice, funding comes from several different sources, cannot be mixed, and is in



various stages of uncertainty. Finally, as in construction work, there is a scheduling problem. We may need to hire two or three contractors at once, because one may not be able to simultaneously work on three projects which must be done by a certain date.

In practice then, a person with, say three, projects would probably

1. do none of them
2. do them all together, if possible
3. do separately as required by funding and scheduling.

Additional complications may occur in the project selection/capital budgeting problem in that

1. the selection of one activity may either necessitate or preclude the possibility of the selection of another activity
2. the total number of activities is limited
3. the total amount of money is limited (perhaps cancelling some optional activities).

It appears, in view of the difficulties described above, that a new model which takes the scheduling and funding, at least, into account may be developed. (Alexander [1] and Weingartner [29]).

#### h. Political Districting

In political districting, an area (e. g., a state) is partitioned into a number of smaller areas called districts which are assigned a number of representatives. The area has also been partitioned into a number of basic population units (e. g., counties). A political districting plan consists of assigning the population units to the districts so that:

- i) a population unit is assigned to exactly one district
- ii) the total population assigned to a district doesn't deviate from the mean district population by more than a given number
- iii) a district is contiguous, i. e., it is possible to walk between population units making up the district without going into another district (roughly speaking)
- and iv) a district is compact, i. e., somewhat circular or square

Details and examples are given in (Garfinkel and Nemhauser [10]). Without the contiguity and compactness restrictions, this problem may be formulated as a QSCP. Suppose we have  $n$  basic population units  $a_i$  ( $i = 1, 2, \dots, n$ ) with the population of unit  $a_i$  being  $d_i$ , and  $m$  districts  $b_k$  ( $k = 1, 2, \dots, m$ ). Define variables  $x_{ik}$  by

$$x_{ik} = \begin{cases} 1 & \text{if } a_i \text{ is assigned to } b_k \\ 0 & \text{otherwise} \end{cases}$$

and let  $x' = (x_{11} \ x_{12} \ \dots \ x_{1m}; \ x_{21} \ x_{22} \ \dots \ x_{2m}; \ \dots; \ x_{n1} \ x_{n2} \ \dots \ x_{nm})$

The QSCP is then:

$$\begin{aligned} &\text{minimize} && x'Cx \\ &\text{subject to} && Hx = e_n \\ &&& \text{and } x \text{ a binary vector} \end{aligned}$$

where  $C$  is the  $(n \cdot m) \times (n \cdot m)$  partitioned matrix

$$C = \begin{bmatrix} c_{11} & c_{12} & \dots & c_{1m} \\ \hline c_{21} & c_{22} & \dots & c_{2m} \\ \hline \vdots & & & \\ \hline c_{m1} & & & c_{mm} \end{bmatrix}$$

$$C_{ij} = \begin{bmatrix} d_1^2 - 2dd_1 & d_1 d_2 & \dots & d_1 d_n \\ d_2 d_1 & d_2^2 - 2dd_2 & & d_2 d_n \\ \vdots & \vdots & & \vdots \\ d_n d_1 & d_n d_2 & \dots & d_n^2 - 2dd_n \end{bmatrix} \quad \text{if } i = j$$

and  $C_{ij}$  = the  $n \times n$  zero matrix for  $i \neq j$

(here  $d = \frac{1}{m} \sum_{i=1}^n d_i$ , the mean district population)

and  $H$  is the  $n \times (n \cdot m)$  matrix

$$H = \begin{bmatrix} e_m' & & & \\ & e_m' & & \\ & & \ddots & \\ & & & e_m' \end{bmatrix}$$

## II.5 Miscellaneous Applications

The coloring problem (Bessiere [6]), determining the minimum number of colors needed to color a map so that no two adjacent areas have

the same color, is a LSPP. (Lawler [19]) discusses the problem of minimizing "latency" in magnetic drum on disc storage computers as a QAP. The reader may see applications to staff and class scheduling, constrained least squares estimation, or message switching centers.

## II.6 Some Comments on the Practicability of QSCP Applications

In this section we discuss the realistic values of the applications given in sections II.2, II.3, and II.4. Quite often, simple examples are used to illustrate the application of some theory, and, while they serve the purpose of clarifying the theory, they are of no practical use. Basically there are three reasons for this:

- i) the real problem is too simple to waste time applying a complicated theory
- ii) the theory is not complicated enough to model the real problem
- iii) the theory is OK for the real problem but leads to computationally infeasible problems.

We expect the first case (i) to hold for many facilities location and building layout problems. For example, a motel is a relatively uncomplicated structure and probably wouldn't require a QAP to determine a layout. The true value of the QAP is apparent when a complicated structure such as a hospital is designed or a complex wiring problem is considered. Unfortunately, in these cases the problem may become computationally infeasible (iii).

A good example where (ii) the theory was not complicated enough is the project selection problem of II.4.g. Many of the references cited state that computationally infeasible problems result from the application of the theory described to real problems: (Jarvis [18, p. 88]) states

that although the attacker problem can be handled in a reasonably efficient manner, the defender problem cannot. Similar statements occur in (Maxwell [21]) concerning production scheduling, in (Gorenstein [14]) on printing press scheduling, and in (Freund and Sadosky [9]) on control panel layout. In other words, there are a large number of valuable applications, once efficient solution procedures are developed. These procedures may include new algorithms and combinations/modifications of old ones, or the attempt to include more complications of the original problem as in the GPP.

How about the use of quadratic versus linear models for some applications? As was mentioned in section II.4 some problems, e. g., the information retrieval problem, are naturally linear and no basis was found to consider quadratic interaction. Examples of applications of set covering problems with quadratic objective functions, with the exception of the special case of the QAP, are not easy to find. Most examples considered, which were not linear, such as construction project selection, exhibit higher order interactions. It does not appear that examples with high order interactions can be approximated by quadratic interactions in a successful manner. Although (Weingartner [29]) points out the tremendous advantage of the quadratic over the linear model for project selection/capital budgeting, some concrete, real applications remain to be found which can be solved as, say, a QSPP.

## REFERENCES

1. Alexander, D., Interview, Physical Plant, Georgia Institute of Technology, 1973.
2. Arabeyre, J. P., J. Fearnley, F. C. Steiger, and W. Teather, "The Airline Crew Scheduling Problem: A Survey," *Transportation Science* 3, 140-163, 1969.
3. Balinski, M. L. and R. E. Quandt, "On an Integer Program for a Delivery Problem," *Operations Research* 13, 300-304, 1964.
4. Bellmore, M. and H. D. Ratliff, "Set Covering and Involutory Bases," *Man. Sci.* 18, 194-205, 1971.
5. Bellmore, M. and H. D. Ratliff, "Optimal Defense of Multi-Commodity Networks," *Man. Sci.* 18, 174-185, 1971.
6. Bessiere, F., "Sur la Recherche der Nombre Chromatique d'Un Graph par une Programme Lineaire en Nombres Entiers," *Rev. Franc, Recherche Operat.* 9, 143-148, 1965.
7. Day, K. H., "On Optimal Extracting from a Multiple File Data Storage System: An Application of Integer Programming," *Operations Research* 13, 482-495, 1965.
8. Dorris, A. L., "The Utility of Optimization Techniques in the Design of Man-Machine Systems," Master's Thesis, Georgia Institute of Technology, 1971.
9. Freund, L. E. and T. L. Sadosky, "Linear Programming Applied to Optimization of Instrument Panel and Workplace Layout," *Human Factors* 9, 295-300, 1967.
10. Garfinkel, K. S. and G. L. Nemhauser, "Optimal Political Districting by Implicit Enumeration Techniques," *Man. Sci.* 16, 495-504, 1970.
11. Garfinkel, K. S. and G. L. Nemhauser, Integer Programming, John Wiley & Son, 1972.
12. Garg, K. C., S. Kumar, P. Dass, and P. Len, "Generalized Traveling Salesman Problem Through n Sets of Nodes in a Competitive Market," *Ablanf and Plassungsforschung (Germany)* 11, 116-120, 1970.
13. Geddes, P. H., "The Solution of Large Size Quadratic Set Covering Problems," Master's Thesis (in progress), School of Industrial and Systems Engineering, Georgia Institute of Technology.

14. Gorenstein, S., "Printing Press Scheduling for Multi-Edition Periodicals," *Man. Sci.* 16, 373-383, 1970.
15. Graves, G. W., and A. B. Whinston, "An Algorithm for the Quadratic Assignment Problem," in Integer and Nonlinear Programming, ed. by J. Abadie, North Holland Pub. Co., Amsterdam, 473-497, 1970.
16. Hanan, M., and J. M. Kurtzberg (1972), "A Review of the Placement and Quadratic Assignment Problems," *SIAM Review* 14, 324-342, 1972.
17. Hu, T. C., Integer Programming and Network Flows, Addison-Wesley, 1969.
18. Jarvis, J. J., Optimal Attack and Defense of a Command and Control Communications Network, Ph.D. Dissertation, The Johns Hopkins University, 1968.
19. Lawler, E. D., "Notes on the Quadratic Assignment Problem," Harvard Computation Laboratory, unpublished paper, 1960.
20. Loberman, H., and A. Weinberger, "Formal Procedures for Connecting Terminals with a Minimum Total Wire Length," *Journal of the Association of Computing Machinery* 4, 428-437, 1957.
21. Maxwell, W. L., "The Scheduling of Single Machine Systems: A Review," *The International Journal of Production Research* 3, 177-199, 1964.
22. Murty, K. G., "On the Set Representation and Set Covering Problems," in Symposium on the Theory of Scheduling and its Applications, ed. by M. Beckman, G. Goos, and H. P. Kunzi, Springer-Verlag, 143-163, 1973.
23. Salvesan, M. E., "The Assembly Line Balancing Problem," *Journal of Industrial Engineering* 6, 18-25, 1955.
24. Srivastava, S. S., S. Kumar, K. C. Garg, and P. Len, "Generalized Traveling Salesman Problem through n Sets of Nodes," *CORS Journal* 7, 97-101, 1969.
25. Steinberg, L., "The Backboard Wiring Problem: A Placement Algorithm," *SIAM Review* 3, 37-50, 1961.
26. Svestka, J. A. and V. E. Huckfeld, "Computational Experience with an M-Salesman Traveling Salesman Algorithm," *Man. Sci.* 19, 790-799, 1973.
27. Tillman, F. A., and R. W. Hering, "A Study of a Look-Ahead Procedure for Solving the Multiterminal Delivery Problem," *Transportation Research* 5, 225-229, 1971.
28. Tsoi, A. K. H., S. M. Taskai, and V. N. Druzhinin, "The Scheduling of Movements of Excavators as a Problem of Several Traveling Salesmen," *Trudy Instituta Gormogo Dela, Akademirijo Nauk Kazakhrkogo SSR (U. S. S. R.)* 45, 9-13, 1971.

29. Weingartner, H. M., "Capital Budgeting of Interrelated Projects: Survey and Synthesis," Man. Sci. 12, 458-516, 1966.
30. Whitehead, N. and M. Z. Elders, "An Approach to the Optimum Layout of Single-Story Buildings," Architects' Journal 139, 1373-1380, 1964.
31. Wismer, D. A., "Solution of the Flowshop-Scheduling Problem with no Intermediate Queues," Operations Research 20, 689-697, 1972.
32. Zak, Yu. A., "Solution Algorithms for 'n Traveling Salesmen' Problem," Kibernetika (U. S. S. R.) 1, 99-106, (Russian, English Summary), 1972.



### III. THE QUADRATIC SET COVERING PROBLEM

In this section we will develop a branch and bound scheme for solving quadratic set covering problems which arise in the context of layout design.

#### III.1 Introduction

Consider the following quadratic set covering problem.

Minimize  $x'Cx$

subject to  $Ax \geq e$

$x$  is a binary vector

Several procedures for solving problems with the above structure exist in the literature. In [2] Bazaraa and Goode presented a cutting plane algorithm for solving such problems. The algorithm excludes integer and noninteger points at each step, and stops with an optimal solution after a finite number of steps. The algorithm is suitable for problems with:

- the number of variables  $n \leq 40$
- the number of constraints  $m \leq 60$
- low density of  $A$  and  $C$

In [3], an implicit enumeration algorithm which extends the notions in [1] and [6] to the quadratic case, is developed for the above problem. Problems with up to 100 variables and 100 constraints, and small density of the cost matrix  $C$  can be handled by the procedure. Here, rather than attempting to develop an algorithm for solving any quadratic set covering problem with the above structure, we will develop a procedure for the specific applications of layout design. Existing layout procedures are heuristic in nature and can be classified into the following two categories:

- a. Construction algorithms      The layout is constructed from scratch by successively selecting a new object (department, room, facility,

etc.) until the layout is completed. These algorithms are usually based on qualitative evaluation of the necessity or nonnecessity of adjacency of different objects. CORELAP [8], ALDEP [9], and LSP [11] are examples of such constructive algorithms.

- b. Improvement algorithms      Given an existing layout, an improved layout is sought by interchanging the positions of various objects. The improvement is based on reducing the total flow or cost of flow. CRAFT [4] and the method of Whitehead and Elder [10] are examples of such algorithms.

For more details the reader may refer to Francis and White [5, Chapter 3]. Here we present a quadratic set covering formulation of the layout problem. The formulation handles single or multi-story buildings, objects with regular or irregular shapes, designing a layout from scratch, or adding new objects to an existing layout. The final layout is provided by a branch and bound optimization procedure.

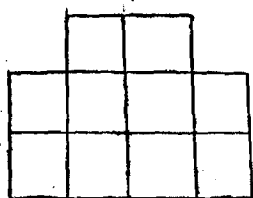
In order to produce an optimal layout the following input data are required. First, the size and required shape (or shapes) of each object must be provided. Secondly, candidate locations of each object are determined. If applicable, a fixed cost of assigning an object to a certain location is given. Finally, the flow (or interaction) between each two objects is determined. If not applicable or hard to estimate, this information may be substituted by the desirability or undesirability of closeness of any two objects.

### III.2 Formulation of the Problem

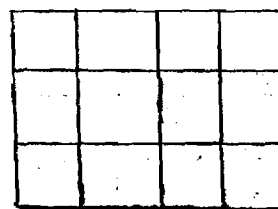
Suppose that we have an area, such as a warehouse, a floor (or multi-floors) of a building, a hospital, etc. Different objects must be

located in the area, such as rooms, manufacturing departments, facilities, etc. An interaction  $u_{ij}$  between objects  $i$  and  $j$  is realized. This interaction may be information flow, material flow, employee movement, etc. The problem is to locate the objects in such a way to minimize the total interaction weighted by the distance, plus any fixed costs that may be realized.

The problem is similar to the quadratic assignment problem, but has some significant differences. In contrast to the quadratic assignment formulation, where each object can occupy a single location, the objects here typically have different areas and designs, as illustrated below.



Object A



Object B

In order to accommodate this situation, the total area is divided into small units or blocks. The possible shapes and areas (in terms of numbers of blocks) of various objects are determined by the layout analyst. The analyst also determines a set of candidate locations for each object.

To illustrate, suppose that the total area is divided in 60 blocks as shown below. The analyst may decide that object A may occupy any one of the following candidate locations:

1. 1, 2, 3, 4, 11, 12, 13, 14, 22, 23
2. 10, 20, 30, 40, 9, 19, 29, 39, 18, 28
3. 55, 56, 57, 58, 45, 46, 47, 48, 36, 37
4. 38, 39, 27, 28, 29, 30, 17, 18, 19, 20
5. 21, 31, 41, 51, 22, 32, 42, 52, 33, 43
6. 14, 15, 23, 24, 25, 26, 33, 34, 35, 36

Each of the above set of blocks constitutes a candidate location of object A. Allowing the analyst to introduce candidate locations for each object helps eliminate undesirable locations\* from the very beginning, takes advantage of the analyst experience, and meanwhile reduces the computational effort since the search space will be reduced.

1	2	3	4	5	6	7	8	9	10
11	12	13	14	15	16	17	18	19	20
21	22	23	24	25	26	27	28	29	30
31	32	33	34	35	36	37	38	39	40
41	42	43	44	45	46	47	48	49	50
51	52	53	54	55	56	57	58	59	60

Let  $x_{ij}$  be 1 if object  $i$  is assigned to its  $j$ th location, and 0 otherwise. Let  $I(i)$  be the total number of possible locations of object  $i$ , and  $J_i(k)$  be the set of blocks occupied by object  $i$  if it were assigned to its  $k$ th location.  $u_{ij}$  denotes the interaction between objects  $i$  and  $j$  and  $f_{ik}$  is the fixed cost (if any) of assigning object  $i$  to its  $k$ th position. The problem is formulated below, where  $d(k_i, l_j)$  is the distance between the centroids of the  $k$ th and  $l$ th locations of objects  $i$  and  $j$ .

\*For example an entrance to the building cannot be located in the interior of the building.

$$\begin{aligned}
\text{Minimize} \quad & \sum_{j=1}^{I(j)} \sum_{k=1}^{I(i)} \sum_{l=1}^m u_{ij} x_{ik} x_{jl} d(k, l_j) + \sum_{k=1}^{I(i)} \sum_{l=1}^m f_{ik} x_{ik} \\
\text{Subject to} \quad & \sum_{k=1}^{I(i)} x_{ik} = 1 \quad \text{for } i = 1, 2, \dots, m \\
& \sum_{k=1}^{I(i)} \sum_{l=1}^m \alpha_{ikt} x_{ik} \leq 1 \quad \text{for each block } t \\
& x_{ik} = 0 \text{ or } 1 \quad \text{for each } k = 1, 2, \dots, I(i) \\
& \quad \text{for each } i = 1, 2, \dots, m
\end{aligned}$$

where  $\alpha_{ikt}$  is 1 if block  $t \in J_i(k)$  and 0 otherwise.

#### Some Comments

- a. The above formulation results in a quadratic set covering problem, where the constraints are of the set covering (partitioning) type and the objective function is quadratic. The first set of constraints requires that each object be assigned to one location and the second set insures that each block is occupied by at most one object.
- b. Since the  $k$ th location of object  $i$  generally consists of many blocks, the distance between locations is taken to be from centroids of the locations. Another method to handle this difficulty is to divide  $u_{ij}$  among the blocks occupied by objects  $i$  and  $j$ , i. e.,  $\bar{u}_{ij} = u_{ij}/I(i) \times I(j)$ . Then the weighted interaction is calculated by shipping  $\bar{u}_{ij}$  units of flow from each block of object  $i$  to each block of object  $j$ , weighted by the distance between the corresponding blocks. For more detailed discussion refer to [10].

- c. The analyst must be careful while estimating  $u_{ij}$ 's, since the flow among different objects may have different units. For example, different types of material are encountered in a manufacturing department, employees with a wide range of salary scale are encountered in a hospital, and so forth. Therefore, a common scale must be devised to obtain uniformity in the estimation of  $u_{ij}$ 's. For more details refer to [10].

In some cases either no flow between objects  $i$  and  $j$  is realized, or else it is difficult to make an accurate estimation of  $u_{ij}$ .

In this case, the analyst may classify the desirability of closeness of any two objects as follows. Given objects  $i$  and  $j$ , the analyst determines if their adjacency or closeness is highly desirable, desirable, immaterial, undesirable, or highly undesirable. A subjective quantification of  $u_{ij}$  for these cases can be made, e. g., 200, 100, 0, -100, and -200 respectively.

- d. Some design restrictions which prohibit certain combinations of assignments can be introduced. To illustrate, the constraint  $x_{12} + x_{27} \leq 1$  would prevent the mutual assignment of the first and second objects to their second and seventh locations, respectively. Such constraints are important in some instances, e. g., objects 1 and 2 may be mutually flammable, and their respective 2nd and 7th locations are adjacent.

### III.3 A Branch and Bound Scheme

We will now describe a branch and bound scheme for finding a layout with minimal objective. At each step of the algorithm a partial layout is at hand where a set of objects are assigned to some locations. A lower bound  $B$  on the cost of all possible completions of this partial layout (solution) is calculated. If  $B$  is less than the cost  $C^*$  of the

best available layout known so far, we proceed by trying to assign a new object. Otherwise the partial solution is fathomed, the last assignment is prohibited, and a new assignment is sought. The method of calculating the lower bound and the progress of the search of the decision tree completely define the procedure.

#### Calculation of the Lower Bound

Suppose that some objects indexed by the set  $I$  have already been assigned. In particular, object  $i$  is assigned to its  $v_i$  possible location. A lower bound  $B$  on the cost of all possible completions of this partial layout is given by  $C_1 + C_2 + C_3$ , where:

$C_1$  = Interaction cost among already assigned objects

$C_2$  = Lower bound on the interaction cost from unassigned objects to assigned objects

$C_3$  = Interaction cost among unassigned objects

#### Calculating $C_1$

$$C_1 = \sum_{\substack{j \in I \\ j \neq i}} \sum_{\substack{l \in I \\ l \neq i}} u_{ij} x_{iv_i} x_{jl} v_j d(v_{i,i}, v_{j,j}) + \sum_{i \in I} f_{iv_i}$$

#### Calculating $C_2 + C_3$

First we find a lower bound of the cost of placing the  $i$ th unassigned object in its  $v$ th location. This is the sum of  $f_{iv}$ , the cost between object  $i$  and the assigned objects, and a lower bound on the cost from other unassigned objects to object  $i$ . The first two costs can be determined exactly.

A lower bound on the cost from other unassigned objects to object  $i$  in its  $v$ th position can be calculated by iteratively finding the position of each unassigned object which minimizes its weighted interaction with object  $i$ , and then adding these minimal weighted interactions. Let the

sum of the above three costs be  $b_{iv}$ . The  $C_2 + C_3$  is found by solving the following problem.

$$\begin{aligned}
 C_2 + C_3 = \text{Minimum} \quad & \sum_{i \notin I} \sum_{v \in A_i} b_{iv} x_{iv} \\
 \text{Subject to} \quad & \sum_{v \in A_i} x_{iv} = 1 \quad i \notin I \\
 & \sum_{i \notin I} \sum_{v \in A_i} \alpha_{ivt} x_{iv} \leq 1 \quad \text{for each unassigned block } t \\
 & x_{iv} \text{ is either 0 or 1}
 \end{aligned}$$

where:

$A_i$  is the collection of available locations of object  $i$  which have not been eliminated by assigning the objects in the set  $I$ , and which have not been eliminated in the branch and bound search

$\alpha_{ivt}$  is 1 if block  $t \in J_i(v)$  and 0 otherwise

The above problem is a linear set covering problem. The first collection of constraints requires that each object be assigned to one of its available locations, and the second set of constraints requires that each vacant block be occupied by at most one unassigned object. The lower bound  $B = C_1 + C_2 + C_3$  is now available. Note that after the above problem is solved, a feasible solution to the overall problem is found and its quadratic cost is evaluated. The best layout  $C^*$  is updated if necessary. Needless to say if the above linear set covering problem admits no feasible solutions, then the current partial solution is fathomed.

There are various ways of simplifying the above problem. For example, we can ignore the second set of constraints. The problem decomposes into a set of trivial knapsack problems, which amounts to independently finding the location of each unassigned object which



minimizes its cost with other objects. This method is adopted here. Another simplification of the above problem is to ignore the integrality restriction of the variables  $x_{iv}$ 's and solving a continuous linear programming problem instead of the linear set covering problem. Of course, these simplifications would generally result in smaller lower bounds thus reducing the speed of fathoming, but on the other hand would result in smaller computational effort per iteration.

#### The Branch and Bound Scheme

We will now describe a branch and bound scheme for finding an optimal layout. We assume familiarity of the reader with general branch and bound schemes. If this were not the case, the reader may refer to (Lawler and Wood [7]).

1. Order the objects, according to the sum of the interactions, say  $C^* = \infty$ . Choose the first object  $i_1$  and place it in a suitable location  $v$ . Let  $A_i$  be the available locations of object  $i$ . Let  $I = \{i_1\}$ ,  $k = 1$ , and go to Step 2.
2. Calculate a lower bound  $B$ , where  $B = \infty$  if no feasible completions exist. If  $B \geq C^*$  go to Step 3. Otherwise go to Step 4.
3. The last assignment of object  $i_k$  to its  $v_k$ th location is prohibited. Try to assign  $i_k$  to another vacant location. If this is possible replace  $v_k$  by this new location and go to Step 2, otherwise free all the locations of object  $i_k$ , remove  $i_k$  from the set  $I$ , and replace  $k$  by  $k - 1$ . If  $k = 0$  go to Step 5. Otherwise repeat Step 3.
4. If  $k = m$  then replace  $C^*$  by  $B$  and store the objects and their corresponding locations as the best layout known so far, and go to Step 3. On the other hand if  $k < m$ ,  $k$  is replaced by  $k + 1$ , and object  $i_k$  is placed in  $I$ . Place  $i_k$  into a vacant location and go to Step 2.

5. Stop. The optimal solution is  $C^*$  and the corresponding layout is optimal.

#### Some Comments

- a. Step 1 is an initialization step. In step 3 the current partial solution is fathomed because  $B \geq C^*$ . If object  $i_k$  could not be assigned anywhere then the level of the tree is reduced by 1 and the current assignment of object  $i_{k-1}$  is prohibited. In Step 4 the incumbent is updated, or else a new object is located.
- b. Given a partial assignment and a new object to be located, there are various methods to determine the location of this object. The location which minimizes the total weighted interaction with the already assigned objects is used here.
- c. Many other rules for choosing a new object to be located could be used. For example, the object with maximum interaction with the already assigned objects may be chosen. A more complicated choice would be the object which if assigned optimally with respect to the already assigned objects, would provide the minimal lower bound of completion.

#### III.4 Computational Experience

In this section we will report our computational experience with two sample problems whose details are given later in the section. The branch and bound technique provided and verified the optimal solutions of both problems.

The first problem involved the assignment of 12 objects with a total area of 53 blocks to an area of 58 blocks. The optimal objective 14079 was obtained in 23.0 seconds on a Univac 1108, and is shown below.

1	2 <b>6</b>	3	4	5	6	7	8 <b>3</b>	9	10
11	12 <b>5</b>	13	14	15 <b>1</b>	16	17	18	19	20
21 <b>12</b>	22	23	24	25	26	27	28	29	30
31	32 <b>7</b>	33	34	35 <b>9</b>	36	37 <b>4</b>	38	39 <b>8</b>	40
41	42	43 <b>2</b>	44	45	46	47	48	49	50
51 <b>10</b>	52	53	54	55	56	57 <b>11</b>	58		

The second problem involved the assignment of 14 objects which occupy an area of 61 blocks, to an area of 63 blocks. The optimal solution with objective 8170.5 was obtained in 4.4 seconds on a Univac 1108 and is depicted below.

1	2 <b>7</b>	3	4	5	6 <b>9</b>	7	8	9
10 <b>1</b>	11	12	13 <b>5</b>	14	15	16	17 <b>6</b>	18 <b>10</b>
19	20	21	22	23	24 <b>8</b>	25	26	27
28 <b>12</b>	29	30 <b>13</b>	31 <b>8</b>	32 <b>14</b>	33 <b>3</b>	34	35	36
37	38	39	40	41	42	43	44 <b>4</b>	45
46 <b>3</b>	47	48	49 <b>7</b>	50	51	52	53	54
55	56	57	58 <b>11</b>	59	60	61	62	63

Needless to say, a post-optimality analysis is needed where other factors which have not been explicitly considered can be introduced, and where minor modifications of the layout are made. To illustrate, block 34 was vacant in the first problem. A decision must be made regarding what to do with this block, e. g., shifting object 7 such that it occupies blocks 32, 33, and 34; or enlarging object 9 so that it occupies blocks 34, 35, and 36, etc.

An alternative set of suboptimal layout designs can also be provided by the branch and bound scheme. Because there are usually some factors which cannot be introduced into the model, the layout engineer may prefer

preferred by the layout engineer.

1	6	2	3	4	5	6	7	8	9	10			
11		12	5	13	14	15	7	16	17	18	19	3	20
21	12	22		23	24	25	26	27	28		29		30
31		32	33	34	11	35	36	37	2	38	39	8	40
41		42	43	44		45	46	47	2	48	49	8	50
51	4	52	53	9	54	10	55	56	11	57	58		

## Problem Statements

The data for the two problems of this section are summarized below. The total interaction or flow between each two objects and the candidate locations of each object in terms of blocks are listed.

### Problem 1

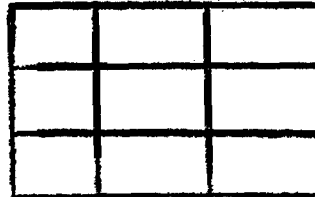
### Interaction Matrix

[illegible]

## Candidate Locations

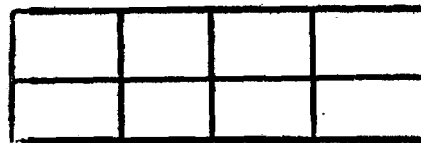
### Object 1

1, 2, 3, 11, 12, 13, 21, 22, 23  
24, 25, 26, 34, 35, 36, 44, 45, 46  
28, 29, 30, 38, 39, 40, 48, 49, 50  
31, 32, 33, 41, 42, 43, 51, 52, 53  
4, 5, 6, 14, 15, 16, 24, 25, 26  
27, 28, 29, 37, 38, 39, 47, 48, 49



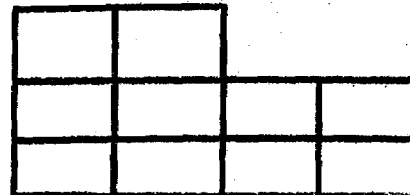
### Object 2

31, 32, 33, 34, 41, 42, 43, 44  
45, 46, 47, 48, 55, 56, 57, 58  
19, 20, 29, 30, 39, 40, 49, 50  
11, 12, 13, 14, 21, 22, 23, 24  
43, 44, 45, 46, 53, 54, 55, 56  
27, 37, 47, 57, 28, 38, 48, 58



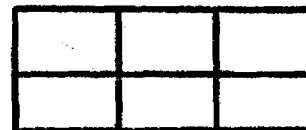
### Object 3

7, 8, 9, 10, 17, 18, 19, 20, 29, 30  
37, 38, 39, 40, 47, 48, 49, 50, 57, 58  
5, 6, 15, 16, 17, 18, 25, 26, 27, 28  
8, 9, 10, 18, 19, 20, 29, 30, 39, 40



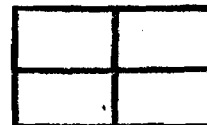
### Object 4

4, 14, 24, 5, 15, 25  
1, 2, 3, 11, 12, 13  
31, 32, 41, 42, 51, 52  
14, 15, 16, 24, 25, 26  
27, 28, 37, 38, 47, 48



### Object 5

45, 46, 55, 56      17, 18, 7, 8  
39, 40, 49, 50      4, 5, 14, 15  
24, 25, 34, 35      12, 13, 22, 23  
22, 23, 12, 13



### Object 6

37, 47, 57      18, 19, 20  
1, 2, 3      8, 9, 10  
30, 40, 50      13, 23, 33  
45, 46, 47



Object 7

48, 49, 50      10, 20, 30  
31, 32, 33      14, 15, 16  
29, 39, 49      16, 17, 18



Object 8

16, 17, 26, 27      19, 20, 29, 30  
39, 40, 49, 50      33, 34, 43, 44  
6, 7, 16, 17      41, 42, 51, 52



Object 9

57, 58      51, 52      42, 43      8, 9  
37, 38      9, 10      43, 53      35, 36  
38, 48      25, 26      1, 11  
29, 30      46, 47      13, 14  
11, 12      41, 42      41, 31



Object 10

52, 53      7, 17      38, 39  
51, 52      49, 50      39, 40  
54, 55      1, 2      42, 43



Object 11

any of 41, 42, 43, ..., 58



Object 12

any of 7, 8, 9, 10, 17, 18, 19, 21, 22, 32



## Problem 2

### Interaction Matrix

	1	2	3	4	5	6	7	8	9	10	11	12	13	14
1	-	72	162	90	108	27	0	0	18	27	18	0	0	0
2		-	72	80	0	48	0	48	32	0	16	8	0	0
3			-	45	54	27	27	27	0	27	0	9	18	0
4				-	30	0	30	30	20	0	20	10	10	0
5					-	18	0	18	12	18	24	0	0	0
6						-	9	9	0	0	6	6	6	0
7							-	9	12	9	6	3	0	0
8								-	6	9	0	3	0	0
9									-	6	4	6	2	0
10										-	6	3	6	0
11											-	2	0	0
12												-	4	0
13													-	0
14														-

### Object 1

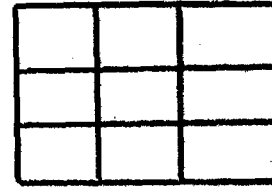
1, 2, 3, 10, 11, 12, 19, 20, 21  
 14, 15, 16, 23, 24, 25, 32, 33, 34  
 37, 38, 39, 46, 47, 48, 55, 56, 57  
 7, 8, 9, 16, 17, 18, 25, 26, 27  
 40, 41, 42, 49, 50, 51, 58, 59, 60  
 6, 7, 8, 15, 16, 17, 24, 25, 26  
 30, 31, 32, 39, 40, 41, 48, 49, 50


### Object 2

13, 4, 22, 31, 5, 14, 23, 32  
 6, 7, 8, 9, 15, 16, 17, 18  
 28, 29, 37, 38, 46, 47, 55, 56  
 39, 40, 41, 42, 48, 49, 50, 51  
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 15, 16, 24, 25, 33, 34, 42, 43  
 22, 23, 24, 25, 31, 32, 33, 34

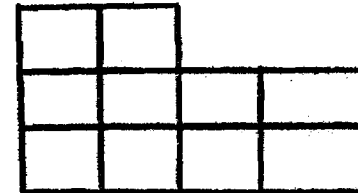

Object 3

43, 44, 45, 52, 53, 54, 61, 62, 63  
37, 38, 39, 46, 47, 48, 55, 56, 57  
13, 14, 15, 22, 23, 24, 31, 32, 33  
39, 40, 41, 48, 49, 50, 57, 58, 59  
7, 8, 9, 16, 17, 18, 25, 26, 27  
33, 34, 35, 42, 43, 44, 51, 52, 53



Object 4

29, 30, 38, 39, 40, 41, 47, 48, 49, 50  
35, 36, 44, 45, 52, 53, 54, 61, 62, 63  
1, 2, 3, 4, 10, 11, 12, 13, 19, 20  
21, 30, 4, 13, 22, 31, 5, 14, 23, 32  
42, 51, 60, 43, 52, 61, 53, 62, 54, 63  
6, 15, 24, 33, 7, 16, 25, 34, 36, 35



Object 5

15, 16, 17, 24, 25, 26  
37, 38, 46, 47, 55, 56  
1, 2, 3, 10, 11, 12  
31, 32, 33, 40, 41, 42  
6, 7, 15, 16, 24, 25  
4, 13, 22, 5, 14, 23



Object 6

34, 35, 36      1, 10, 19  
45, 54, 63      13, 14, 15  
58, 59, 60      8, 17, 26  
55, 56, 57



Object 7

42, 41, 60      49, 50, 51  
4, 5, 6      3, 12, 21  
9, 18, 27      38, 39, 40  
40, 41, 42



Object 8

8, 9, 18      31, 40, 39  
46, 55, 56      43, 44, 53  
22, 23, 32      62, 63, 54  
31, 40, 41



Object 9

6, 7      11, 12      50, 51  
8, 9      1, 10      52, 61  
34, 35      27, 36      62, 63





Object 10

57, 58, 59      3, 4, 5  
45, 54, 63      5, 14, 23  
7, 8, 9      34, 35, 36  
9, 18, 27

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Object 11

27, 36      22, 23  
58, 59      20, 29  
11, 20      19, 28  
5, 6      40, 41

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Object 12

any of 56, 9, 28, 31, 37, 63, 45, 39, 29

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Object 13

any of 28, 46, 37, 56, 7, 24, 30, 19, 45

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Object 14

any of 1, 2, ..., 63

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## REFERENCES

1. Balas, E., "An Additive Algorithm for Solving Linear Programs with Zero-One Variables, Operations Research, Vol. 13, pp. 517-546, 1965.
2. Bazaraa, M. S., and J. J. Goode, "A Cutting Plane Algorithm for the Quadratic Set Covering Problem," Operations Research, to appear.
3. Bazaraa, M. S., R. L. Rardin, and C. M. Shetty, "Solving Quadratic Set Covering Problems by Implicit Enumeration," School of Industrial and Systems Engineering, Georgia Institute of Technology, Atlanta, Georgia.
4. Buffa, E. S., G. C. Armour, and T. E. Vollman, "Allocating Facilities with CRAFT," Harvard Business Review, Vol. 42, pp. 136-159, 1964.
5. Francis, R. L. and J. A. White, "Facility Layout and Location: An Analytical Approach," Prentice-Hall, 1974.
6. Geoffrion, A. M., "An Improved Implicit Enumeration Approach for Integer Programming," Operations Research, Vol. 17, 1969.
7. Lawler, E. L. and D. E. Wood, "Branch and Bound Methods: A Survey," Operations Research, Vol. 14, pp. 699-719, 1966.
8. Lee, R. C., and J. M. Moore, "CORELAP--Computerized Relationship Layout Planning," Journal of Industrial Engineering, Vol. 18, pp. 195-200, 1967.
9. Sheehof, J. M., and W. O. Evans, "Automatic Layout Design Program," The Journal of Industrial Engineering, Vol. 18, pp. 690-695, 1967.
10. Whitehead, N., and M. Z. Elders, "An Approach to the Optimum Layout of Single Story Buildings," Architect Journal, Vol. 139, pp. 1373-1380, 1964.
11. Zoller, K., and K. Adendorff, "Layout Planning by Computer Simulation," AIIE Transactions, Vol. 4, pp. 116-125, 1972.

#### IV. THE QUADRATIC ASSIGNMENT PROBLEM

In this section an exact branch and bound procedure which is able to produce optimal solutions for problems with twelve facilities or less, is developed. A new formulation of the problem and a duality based sub-gradient algorithm are pretested. Limited computational experience shows some difficulties with the latter approach relative to the size of the duality gap. Since the exact procedures were not satisfactory for large problems, a heuristic is developed which was able to produce good quality solutions with a reasonable computational effort. Computational experience with the exact and heuristic codes is presented.

##### IV.1 Introduction

Recall that the quadratic set covering problem  $P_1$  can be formulated as follows:

$$\begin{array}{ll}
 \text{Minimize} & \sum_{\ell=1}^n \sum_{k=1}^m \sum_{j=1}^n \sum_{i=1}^m c_{ijkl} x_{ij} x_{k\ell} + \sum_{j=1}^n \sum_{i=1}^m f_{ij} x_{ij} \\
 \text{s.t.} & \sum_{j=1}^n x_{ij} = 1 \quad i = 1, 2, \dots, m \\
 & \sum_{i=1}^m x_{ij} \leq 1 \quad j = 1, 2, \dots, n \\
 & x_{ij} \text{ is 0 or 1} \quad i = 1, 2, \dots, m, j = 1, 2, \dots, n
 \end{array}$$

The following interpretation will be used here (see II.1 and II.2).

We suppose that  $m$  objects are to be assigned to  $n$  locations ( $n \geq m$ ).  $x_{ij}$  is 1 if object  $i$  is placed in location  $j$  and is 0 otherwise.  $c_{ijkl}$  is the cost of mutually assigning object  $i$  to location  $j$  and object  $k$  to location  $\ell$ .

$c_{ijkl}$  is usually determined as the number of interactions  $u_{ik}$  between objects  $i$  and  $k$  weighted by the distance from location  $j$  to location  $\ell$ , i. e.,

$c_{ijkl} = u_{ik} d_{j\ell}$ . In some instances a fixed cost related to the mutual

assignments  $x_{ij}$  and  $x_{k\ell}$  may be incurred, say  $\alpha_{ijkl}$  (See Graves and Whinston

[10]), in which case  $c_{ijkl} = u_{ik}d_{jl} + \alpha_{ijkl} \cdot f_{ij}$  is the fixed cost of assigning object  $i$  to location  $j$ .

As discussed in II.2, applications of the problem range from location of indivisible facilities (Koopmans and Beckman [17]), building layout (Whitehead and Elders [26]), control panel layout (Dorris [5]), to wiring design in the placement of electronic components in assemblies (Breuer [3], Gaschutz and Ahrens [7], and Steinberg [24]).

Various procedures for solving the problem have been suggested in the literature, including both exact and heuristic methods. Exact methods include the integer programming formulation of (Lawler [19]) and several branch and bound procedures. The latter class of methods include the single assignment procedures of (Gilmore [9] and Lawler [17]) and the double assignment methods of (Land [18] and Gavett and Plyter [8]). These methods have either not been tested or else found impractical with  $n$  larger than 7 (See Pierce and Crowston [23]).

Several heuristics have been reported in the literature. The reader may refer to the procedures of (Armour and Buffa [1], Gaschutz and Ahrens [7], Gilmore [9], Graves and Whinston [10], Heider [12], Hillier [15], Hillier and Connors [16], Neghabat [20], Nugent, Vollman, and Ruml [21], Pegels [23], Steinberg [24], Whitehead and Elders [26], Vollman, Nugent and Zartler [25]). The reader may also refer to the papers of (Hannan and Kurtzberg [11]) and (Pierce and Crowston [23]), as well as the book of (Francis and White [6, Chapter 8]) which compare among various methods of solving the quadratic assignment problem.

In IV.2 we present an exact branch and bound procedure, which can be used to obtain optimal and suboptimal solutions with any apriori desired degree of accuracy. Of course the solution time increases as a function

of the accuracy desired. The procedure was able to optimally solve problems with size up to 12 facilities. The problem of adding new facilities to an existing layout is also discussed. In IV.3 a reformulation of the problem based on double assignment was devised. A duality based method is suggested to solve the new formulation. The approach did not prove particularly successful from a computational point of view for several reasons which are discussed. A heuristic procedure is reported in IV.4. Computational experience with both exact and heuristic codes is reported in IV.5.

#### IV.2 An Exact Branch and Bound Procedure

A branch and bound solution procedure which can be used to obtain optimal and "qualified" suboptimal solutions will be described here. The following notation will be used. The location to which object  $i$  is assigned is denoted by  $\ell(i)$ . Hence the mutual cost of assigning object  $i$  and  $j$  is  $(u_{ij} + u_{ji}) d_{\ell(i)\ell(j)}$  (assuming that  $d_{\ell(i)\ell(j)} = d_{\ell(j)\ell(i)}$ ) and the fixed costs of these assignments (if any) are  $f_{i\ell(i)} + f_{j\ell(j)}$ .

#### Decision Tree and General Framework

At each stage of the algorithm, we have a set of objects that has already been assigned to certain locations. These already assigned objects form a partial solution of the assignment problem. In order to obtain a feasible solution, i. e., a complete assignment, we must find a completion of the partial solution. Rather than considering all the possible ways of completing the partial solution, we first investigate whether the partial solution may lead to a complete solution, with an objective value which is smaller than the best complete solution that we already have. This is done by calculating a lower bound on the cost of the partial solution plus the cost of completing this partial solution.

Let  $B$  be the lower bound and let  $C^*$  be the cost of the best available assignment. If  $B \geq C^*$  then any completion of the partial solution can lead to no improvement. If this is the case, the partial solution is said to be fathomed, and is abandoned. On the other hand, if  $B < C^*$ , it is worthwhile to pursue the partial solution, by seeking to assign more objects.

#### Calculation of the Lower Bound

Suppose that a set of objects in the set  $I$  have already been assigned to a set of locations  $J$ . In particular, suppose that object  $i$  is assigned to location  $\ell(i)$ . The cost of this partial solution,  $C_1$ , can be determined exactly as follows:

$$C_1 = \sum_{i \in I} f_{i\ell(i)} + \sum_{\substack{j \in I \\ j \neq i}} \sum_{i \in I} u_{ij} d_{\ell(i)\ell(j)}$$

The cost of completing this partial solution consists of the fixed costs of assigning objects to locations, plus two types of interaction costs.

The interaction from unassigned objects to assigned objects, with bound  $C_2$ , plus the interaction between the unassigned objects themselves, with bound  $C_3$ .

$C_2$ : bound on cost from unassigned objects to assigned objects.

$C_2$  is the optimal cost of the following linear assignment problem.

$$\text{Minimize} \quad \sum_{j \notin J} \sum_{i \notin I} b_{ij} x_{ij}$$

$$\text{Subject to} \quad \sum_{j \notin J} x_{ij} = 1 \quad \text{for each } i \notin I$$

$$\sum_{i \notin I} x_{ij} \leq 1 \quad \text{for each } j \notin J$$

$$x_{ij} \geq 0 \quad \text{for each } i \notin I, j \notin J$$

where  $b_{ij}$  is a bound on the cost resulting from assigning the unassigned object  $i$  to the unassigned location  $j$ , e. g.,

$$b_{ij} = f_{ij} + \sum_{t \in J} u_{it} d_{jl}(t) + \sum_{t \in J} u_{ti} d_{lj}(t)$$

Of course, a complete solution of the linear assignment problem can be replaced by the simpler task of reducing the matrix  $(b_{ij})$  such that it has at least one zero in each row and each column by subtracting the minima of the rows from the rows, and the minima of the columns from the resulting columns.  $C_2$  can then be replaced by the sum of the row minima plus the sum of the smallest  $m - |I|$  column minima, where  $|I|$  is the number of already assigned objects. Of course, this calculation is simpler, but the bound is not as tight as the bound obtained from solving the assignment problem. Both of these methods for calculating  $C_2$  are used in the computational experience section.

$C_3$ : bound on cost between unassigned objects and themselves.

#### Strategy 1

In order to calculate  $C_3$ , rank the interactions  $u_{ij}$  in a descending order for  $i, j \notin I$ , and rank the distances  $d_{ij}$  in an ascending order for  $i, j \notin J$ . This results in two vectors of size  $(m - |I|)^2$  and  $(n - |J|)^2$ . If  $n > m$ , complete the size of the first vector such that it has  $(n - |J|)^2$  entries by adding zeros to it from the bottom. Then  $C_3$  is the inner product of the two vectors. In other words,  $C_3$  is calculated by matching the largest interaction among unassigned elements to the smallest distance between unassigned locations, and 2nd largest interaction to 2nd smallest distance, and so forth. Clearly, this procedure will give a lower bound on the cost among unassigned elements and themselves.

## Strategy 2

A bound on the cost of interaction among unassigned objects can be alternatively found by solving a linear assignment problem whose cost matrix is constructed as follows. For each unlocated element  $i$ , rank the interactions between it and all other unlocated elements in a descending order. Similarly, for each vacant location  $j$ , rank the distances between it and all other vacant locations in an ascending order. To calculate the entry  $e_{ij}$  find the inner product of the above two vectors.  $C_3$  is found by solving the following linear assignment problem:

$$\begin{array}{ll} \text{Minimize} & \sum_{j \notin J} \sum_{i \notin I} e_{ij} x_{ij} \\ \text{Subject to} & \sum_{j \notin J} x_{ij} = 1 \quad \text{for each } i \notin I \\ & \sum_{i \notin I} x_{ij} \leq 1 \quad \text{for each } j \notin J \\ & x_{ij} \geq 0 \quad \text{for each } i \notin I, j \notin J \end{array}$$

Needless to say, the above assignment problem can be combined with the assignment problem in the  $C_2$  calculation to give  $C_2 + C_3$ . The overall lower bound  $B = C_1 + C_2 + C_3$  is now calculated.

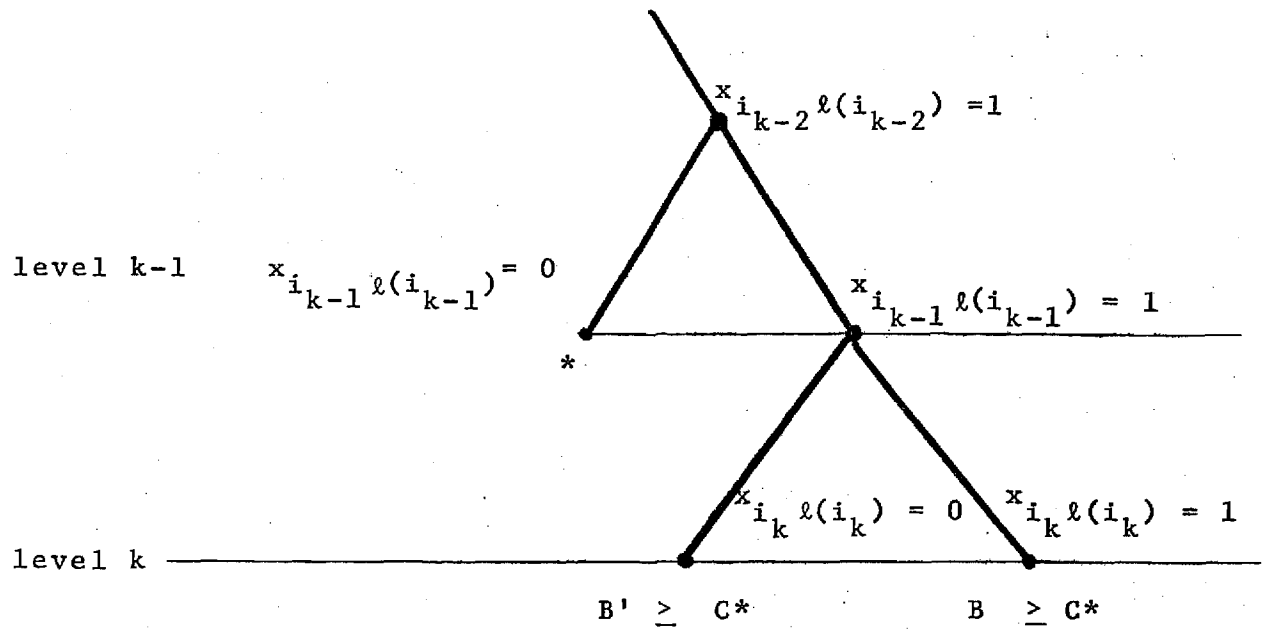
## Continuation of the Search

### Fathoming (Backward Move)

Suppose that  $k$  objects indexed by the set  $I$  have already been assigned to  $k$  locations indexed by the set  $J$ . The level of the trees is called  $k$ . A bound on the cost resulting from all completions of the current partial solution  $B$  is calculated as discussed above. If  $B \geq C^*$ , where  $C^*$  is the current best known cost of a complete assignment, then the partial solution is fathomed. The last assignment, i. e., the  $k$ th assignment is

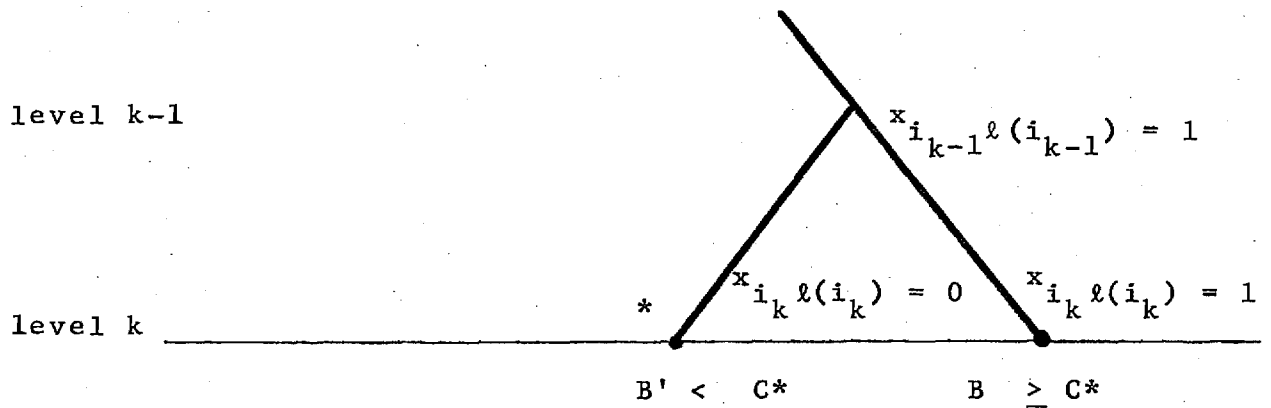


banned, or prohibited, in the hope that this will lead to an improved completion. For example, if the  $k$ th assignment involved placing object  $i_k$  in location  $\ell(i_k)$ , then  $x_{i_k \ell(i_k)}$  is forced to be zero.  $i_k$  is then placed in the unassigned list of objects, (i. e.,  $i_k$  removed from  $I$ ), and similarly  $\ell(i_k)$  is removed from the unassigned list of locations (i. e.,  $\ell(i_k)$  is removed from  $J$ ). A new bound  $B'$  is calculated, in exactly the same manner as explained above, except of course, that the assignment  $x_{i_k \ell(i_k)} = 1$  is prohibited, say by forcing  $b_{i_k \ell(i_k)} = +\infty$  while solving the linear assignment problem. If  $B'$  is still  $\geq C^*$ , then the partial solution of the first  $k-1$  assignments, while banning the assignment  $i_k$  to  $\ell(i_k)$ , can still lead to no improved solutions. Since the first  $k-1$  assignments with  $x_{i_k \ell(i_k)} = 1$  and  $x_{i_k \ell(i_k)} = 0$  lead to no improvement, then all the possibilities at level  $k$  have been exhausted, and prohibiting the assignment at level  $k-1$  is now possible. This is called strong fathoming, and the level of the tree is thus reduced by 1 unit, by prohibiting the assignment at level  $k-1$ , and the process is repeated. If on the other hand, the bound  $B'$  is less than  $C^*$ , which will be referred to as weak fathoming, then object  $i_k$  is assigned to some other unassigned location. This is discussed in more detail in the forward move of the search. The cases of strong and weak fathoming are depicted below.



### Strong Fathoming

\* Continue at level k-1 by trying to assign object  $i_{k-1}$  to some unassigned location  $\neq \ell(i_{k-1})$ .



### Weak Fathoming

\* Continue at level k by trying to assign object  $i_k$  to some unassigned location  $\neq \ell(i_k)$ .

### Progress (Forward Move)

If  $B < C^*$  we must choose an object  $i_{k+1}$  for assignment. For example, we may choose  $i_{k+1}$  to be an unassigned object with maximum interactions with already assigned objects, or choose  $i_{k+1}$  to be an unassigned object with maximum interactions with the most recently assigned object  $i_k$ . This object is assigned to an unassigned location  $\ell(i_{k+1})$  which is not prohibited. This location  $\ell(i_{k+1})$  can be picked in such a way that the total weighted interaction with assigned objects is minimal, e. g., choose  $\ell(i_{k+1})$  which minimizes 
$$\sum_{j=1}^k u_{i_{k+1}i_j} d_{t\ell(i_j)} + \sum_{j=1}^k u_{i_ji_{k+1}} d_{\ell(i_j)t} \text{ over } t \notin J \text{ such that } x_{i_{k+1}t} = 1 \text{ is not prohibited.}$$

Choosing an object  $i_{k+1}$  and a location  $\ell(i_{k+1})$  in the above fashion, increases the level of the tree by 1. A bound on completing the partial solution is calculated as before and the process is repeated.

Needless to say, when the level of the tree is  $m$ , and if the cost is less than  $C^*$ , then  $C^*$  is replaced by the cost, and the corresponding assignment is stored.

### Termination

We have described forward and backward progress of the tree search. If the level of the tree ever reaches value zero, then we stop. This would mean that we are currently at level one, and are trying to backtrack. This means that all possible assignments under  $x_{i_1\ell(i_1)} = 1$  and  $x_{i_1\ell(i_1)} = 0$  have already been enumerated, i. e., all possible ways of assigning the  $m$  objects are enumerated, and we stop. The stored assignment and corresponding  $C^*$  give the optimal solution.

### Summary of the Algorithm

We have discussed above all the details required to describe the following

solution procedure of the quadratic assignment problem.

### Initialization Step

Let  $P(i) = \emptyset$  for  $i = 1, 2, \dots, m$  (prohibited locations). Let  $C^* = \infty$ . Choose an object  $i_1$  and place it in location  $\ell(i_1)$ .  $i_1$  can be determined by maximizing  $\sum_{j=1}^m u_{ij}$  for  $i = 1, 2, \dots, m$ , and  $\ell(i_1)$  can be determined by minimizing  $\sum_{i=1}^n d_{ij}$  for  $j = 1, 2, \dots, n$ . Let  $I = \{i_1\}$  and  $J = \{\ell(i_1)\}$ . Let  $k = 1$  and go to step 1.

### Step 1 (Forward Move)

Calculate a lower bound  $B$  on all completions of current partial solution.  $B = C_1 + C_2 + C_3$ , where  $C_1$ ,  $C_2$ , and  $C_3$  are calculated as shown earlier, with the exception that  $b_{ij} = \infty$  if  $j \in P(i)$ . If  $B \geq C^*$  go to step 2.

Otherwise pick  $i_{k+1} \notin I$  such that  $u_{i_{k+1}i_k} = \max_{i \notin I} u_{ii_k}$ , and place  $i_{k+1}$  in  $\ell(i_{k+1})$  where  $\ell(i_{k+1})$  is determined by

$$\text{Minimizing } f_{i_{k+1}j} + \sum_{t=1}^k u_{i_{k+1}i_t} d_{j\ell(i_t)} + \sum_{t=1}^k u_{i_t i_{k+1}} d_{\ell(i_t)j} \\ \text{for } j \notin J, j \notin P(i_{k+1})$$

Replace  $I$  by  $I \cup \{i_{k+1}\}$  and  $J$  by  $J \cup \{\ell(i_{k+1})\}$ .

If  $k = m - 1$ , calculate the new value of  $C_1$  (cost of complete assignment).

If  $C_1 < C^*$ , replace  $C^*$  by  $C_1$ , store  $(i_t, \ell(i_t))$  for  $t = 1, 2, \dots, m$ , replace  $k$  by  $m$  and go to step 2. If  $C_1 \geq C^*$  then replace  $k$  by  $m$  and go to step 2.

If  $k < m - 1$ , then replace  $k$  by  $k + 1$  and repeat step 1.

### Step 2 (Fathom)

$i_k$  is removed from  $I$  and  $\ell(i_k)$  is removed from  $J$ . Replace  $P(i_k)$  by  $P(i_k) \cup \{\ell(i_k)\}$ . Calculate a lower bound  $B$  on completions of the partial solution  $x_{i_t \ell(i_t)} = 1$  for  $t = 1, 2, \dots, k-1$  and  $x_{i_k \ell(i_k)} = 0$ , where  $B = C_1 + C_2 + C_3$ , and  $b_{ij} = +\infty$  if  $j \in P(i)$ . If  $B \geq C^*$  go to step 3. Otherwise assign  $i_k$  to a location  $t \notin J \cup P(i_k)$  such that the cost

$f_{i_k t} + \sum_{j=1}^{k-1} u_{i_k i_j} d_{t \ell(i_j)} + \sum_{j=1}^{k-1} u_{i_j i_k} d_{\ell(i_j) t}$  is minimized over  
 $t \notin J \cup P(i_k)$ .  $i_k$  is added to  $I$ , and  $t$ , the new  $\ell(i_k)$ , is added to  $J$ .  
 Go to step 1.

### Step 3 (Strong Fathom)

$i_{k-1}$  is deleted from  $I$  and  $\ell(i_{k-1})$  is deleted from  $J$ .  $\ell(i_{k-1})$  is  
 placed in  $P(i_{k-1})$  and  $P(i_k)$  is replaced by the empty set.  $k$  is replaced  
 by  $k-1$ . If  $k = 0$  go to step 4, otherwise go to step 2.

### Step 4 (Termination)

Search has been completed. Optimal cost is  $C^*$  and its corresponding  
 assignment  $(i_k, \ell(i_k))$ ,  $k = 1, 2, \dots, m$  is the optimal assignment. Stop.

The following points are helpful to keep in mind.

1. During the initialization step an upper bound  $C^* = \infty$  is used.  
 As the search progresses,  $C^*$  denotes the objective of the best  
 available complete assignment. Note that  $P(i)$  represents the  
 locations that object  $i$  cannot be assigned to. These are ini-  
 tialized by the empty sets.
2. Step 1 represents a forward step, where the level of the tree in-  
 creases by 1 unit. In this case the bound is less than  $C^*$ ,  
 and hence a complete solution with an objective better than  $C^*$   
 is possible.
3. Step 2 is a fathoming step, where  $B \geq C^*$ . In this case the last  
 assignment is prohibited. Immediately a new bound is calculated.  
 If the new bound is less than  $C^*$ , then a forward move is made.  
 But if the bound is still greater than or equal to  $C^*$ , then a  
 strong fathoming (Step 3) is made, and the level of the tree is  
 reduced. Of course, strong fathoming is most desirable, since it

avoids the expensive task of trying to locate object  $i_k$  to a free location other than  $l(i_k)$ .

### Sub-Optimal Solutions

Due to the highly combinatorial nature of the problem, the task of finding and verifying the optimal solution within a reasonable computational time, for problems with  $n > 15$ , is almost impossible. In order to solve larger problems, we must resort to sub-optimal solutions. The branch and bound procedure itself can be used to obtain sub-optimal solutions. We propose two methods for implementing this general strategy.

#### Method 1

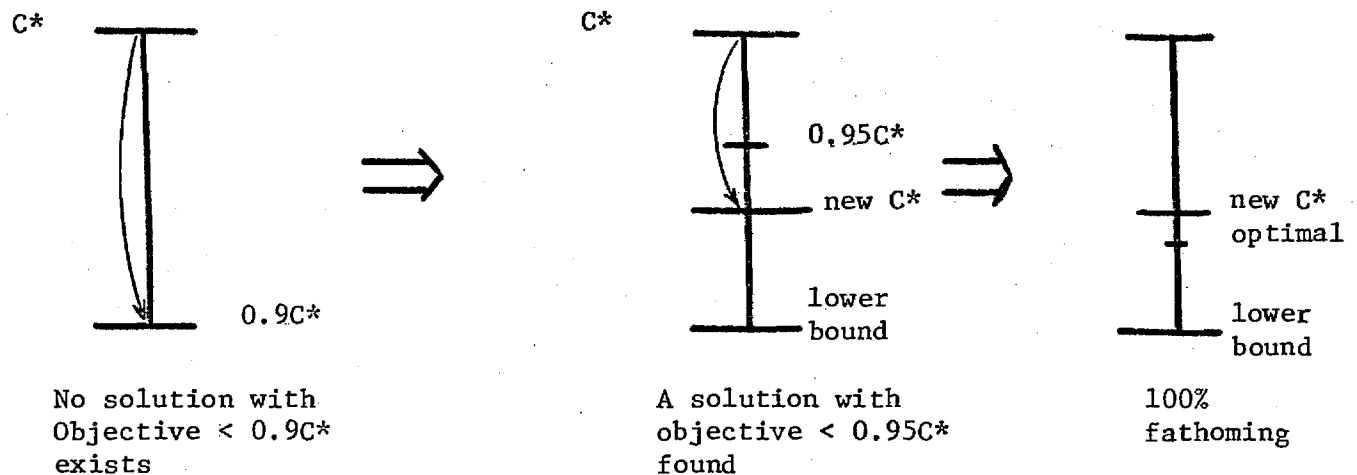
Recall that a partial solution is fathomed, if the lower bound on all completions is at least as big as  $C^*$ , the best known objective. Suppose that a partial solution is fathomed if  $B \geq \alpha C^*$ , where  $\alpha \in [0,1]$ . In this case, the partial solution is abandoned if there is no hope that it leads to an objective which is better than  $\alpha C^*$ . The objective of this simple strategy is clear. We want to fathom, i. e., abandon partial solutions, quickly, even if they lead to a slight improvement. Of course, as a new  $C^*$  is found, then we fathom whenever the bound is greater or equal to  $\alpha$  times the new  $C^*$ . The procedure continues until we cannot find a feasible solution with an objective less than  $\alpha C^*$ . So we have a feasible assignment with objective  $C^*$ , coupled with the statement that the optimal objective is greater than or equal to  $\alpha C^*$ .

#### Choice of $\alpha$

Of course, if  $\alpha$  is small, then fathoming will speed up considerably, resulting in a small computational effort. But on the other hand, the quality of the best obtained feasible solution is not satisfactory. We recommend values of  $\alpha \geq 0.9$  depending on the accuracy required.

### Exact Solution by Stepped Fathoming

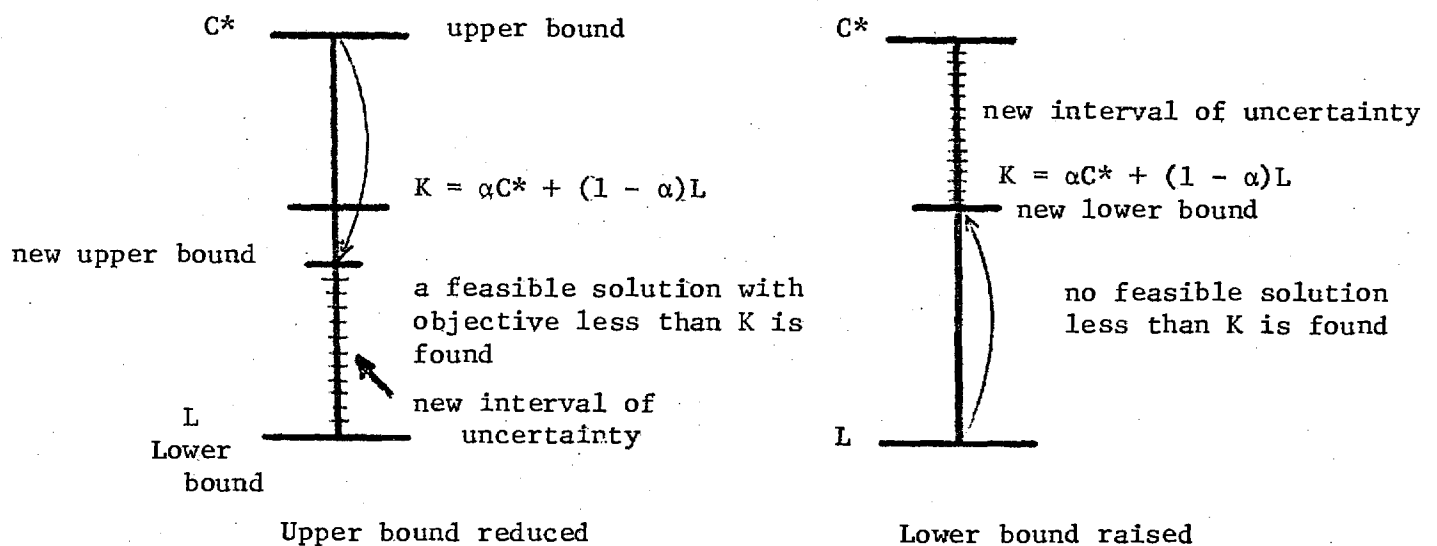
A slight modification of the above scheme can be used to find the exact optimal solution, and still reducing the portion of the tree explicitly enumerated. Suppose that an initial  $\alpha_1$  is used, say  $\alpha_1 = 0.90$ . Eventually we achieve a final  $C^*$  and cannot find a feasible solution with objective less than  $0.90C^*$ , i. e.,  $0.9C^*$  is a lower bound on all solutions. We either stop here, or else switch to  $\alpha_2 \geq \alpha_1$ , say  $\alpha_2 = 0.95$ . We start the search from the stored best solution corresponding to  $C^*$ , which also included all the prohibited locations. We are either able to find a feasible solution with objective less than  $0.95C^*$  (none less than  $0.9C^*$  exist), and repeat the process, or else conclude that none exists. We may then switch to 97% fathoming and eventually to 100% fathoming. This, of course, will lead to the optimal solution. Even though parts of the tree will be searched more than once, the quick fathoming outweighs any repeated effort.



### Method 2

At each stage of the algorithm, we have an upper bound  $C^*$ . A lower bound on the overall problem  $L$ , can be devised. Rather than fathoming on  $C^*$ , suppose we fathom on  $K = \alpha C^* + (1 - \alpha)L$ , where  $\alpha \in [0, 1]$ . Since

$L < C^*$ , then  $K = \alpha C^* + (1 - \alpha)L \leq C^*$ . Two cases are possible. In the first case, we will be able to find a complete solution with objective less than  $K$ . The objective of this new solution becomes the new upper bound  $C^*$  and the process is repeated. In the second case, we will not be able to find such a solution. This automatically implies that there are no solutions with objective less than  $K$ , and hence  $K$  itself is the new lower bound. The process is repeated. From this we keep narrowing the gap between the lower and upper bounds, either by lowering the upper bound (finding an improved feasible assignment) or by raising the lower bound (by finding that there is no feasible solution with objective lower than the current lower bound). When the difference between the lower and upper bounds is smaller than a prescribed tolerance, we either stop, or switch to 100% fathoming. Of course, if the tolerance is zero, or if we switch to 100% fathoming (i. e., switch  $\alpha$  to 1), then we will find the true optimal solution.



#### Choice of $\alpha$

$\alpha$  is any number in the interval  $[0,1]$ . Of course, if  $\alpha$  is close to 1, then we are in effect fathoming on a number very close to  $C^*$ , and the search



will not speed considerably. On the other hand, if  $\alpha$  is close to zero, then improvement is achieved only if we obtain a feasible solution very close to the overall lower bound. In this case, fathoming will be fast, but it is likely not to obtain feasible solutions less than K. So again a tradeoff is required here.

#### Calculation of initial overall lower and upper bounds

Initial lower and upper bounds are needed for implementing the above fathoming scheme. To calculate the lower bound, first calculate a lower bound  $b_{ij}$  on the cost of assigning object  $i$  to location  $j$ , for  $i = 1, 2, \dots, m$  and  $j = 1, 2, \dots, n$ . This is done as discussed earlier. Then a linear assignment problem is solved to find  $L$ . More precisely, let  $L$  be the optimal objective of the following problem.

$$\begin{aligned} \text{Minimize} \quad & \sum_{j=1}^n \sum_{i=1}^m b_{ij} x_{ij} \\ & \sum_{j=1}^n x_{ij} = 1 & i = 1, 2, \dots, m \\ & \sum_{i=1}^m x_{ij} \leq 1 & j = 1, 2, \dots, n \\ & x_{ij} \geq 0 & \text{all } i \text{ and } j \end{aligned}$$

An upper bound  $C^*$  is immediately available, by calculating the quadratic cost of the optimal assignment resulting from the above problem. Now Method 2 can be initiated.

#### Adding New Facilities to an Existing Layout

In many applications a large number of objects is already preassigned, and only some new objects are to be placed, such that the overall cost is minimized. In this case, the above algorithm can be applied with few obvious modifications in the calculations. Since the preassigned objects and their locations will always be assigned, then these objects will always be in the set  $I$  and their locations will always be in the set  $J$ . In the tree search,

if  $\gamma$  objects are already assigned, we start the search by assigning more objects, i. e., the level of the tree starts at  $\gamma + 1$ . If the level of the tree ever becomes  $\gamma$ , then we stop. If an overall lower bound on all feasible solutions to the problem is desired, we recommend calculating  $C_1 + C_2 + C_3$  as before, where I and J are the indices of the preassigned objects and their locations.

It is interesting to note that the problem of assigning  $m$  facilities anew is usually harder to solve than a similar problem where  $m$  new facilities are to be assigned to an existing layout. This is due to the fact that the lower bounds in the latter case are usually tighter, thus allowing quicker fathoming.

#### IV.3. Reformulation of the Problem

At a certain stage, an alternative formulation for the quadratic assignment problem is sought in such a way that an approach based on solving the dual problem would be involved. Under all outcomes, it would either produce a solution to the original problem or at least a set of dual variables which, when appropriately added in the original cost data would result in higher bounds at the stage of applying the tree search scheme. This would be expected to make search faster and hence reduce the computational difficulty of the problem in its known form.

##### Reformulation

Define the following:

$$\begin{aligned}
 c_{ijkl} &= f_{ij} + f_{kl} + (u_{ik} + u_{ki}) d_{jl} \\
 y_{ijkl} &= \begin{cases} 1 & \text{if object } i \text{ is located in } j \text{ and object } k \text{ is located in } l \\ 0 & \text{otherwise} \end{cases} \\
 x_{ij} &= \begin{cases} 1 & \text{if object } i \text{ is located in } j \\ 0 & \text{otherwise} \end{cases}
 \end{aligned}$$

$\alpha_i$  is the number of objects interacting with object  $i$ .

Then the problem could be reformulated as follows:

$$\begin{array}{ll}
 \text{P2:} & \text{Minimize} \quad \sum_i \sum_j \sum_k \sum_l c_{ijkl} y_{ijkl} \\
 & \text{s.t.} \quad \begin{array}{l}
 \text{(a)} \quad \left\{ \begin{array}{l} \sum_k \sum_l y_{ijkl} - \alpha_i x_{ij} = 0 \quad \text{for all } i, j \\ \sum_i \sum_j y_{ijkl} - \alpha_k x_{kl} = 0 \quad \text{for all } k, l \end{array} \right. \\
 \text{(b)} \quad \sum_i \sum_k y_{ijkl} = 1 \quad \text{for all } j, l \\
 \text{(c)} \quad \left\{ \begin{array}{l} \sum_i x_{ij} \leq 1 \quad \text{for all } j \\ \sum_j x_{ij} = 1 \quad \text{for all } i \end{array} \right. \\
 \text{(d)} \quad y_{ijkl} \text{ and } x_{ij} \text{ is 0 or 1 for all } i, j, k, l
 \end{array}
 \end{array}
 \quad \begin{array}{l}
 \text{dual} \\
 \text{variables} \\
 \boxed{u_{ij}} \\
 \boxed{v_{kl}}
 \end{array}$$

The above formulation can be thought of as a pair assignment formulation where the number of variables significantly increase. Note that the constraints are linear, except for, integrality of the variables. The formulation implies that  $y_{ijkl} = 1$  if and only if  $x_{ij} = x_{kl} = 1$ . The constraint set (a) specifies that the number of nonzero  $y_{ijkl}$  has to comply with the number of actual connections to element  $i(k)$  and the fact that it is in location  $j(l)$  or not. The set (b) specifies that  $y_{ijkl}$  has to comply with the fact that one element only can be in a location. Constraint set (c) is the standard assignment problem constraint set (i. e., each location can have at most one element and each element can occupy only one location). The 0-1 condition is expressed in the set (d).

Problem P2 can be rewritten as follows, where for the purpose of simplification of notation  $n$  is assumed equal to  $m$ , the dual vectors  $u$  and  $v$  are combined as  $w$ , and  $\alpha$  is a diagonal matrix with obvious entries.

P3:	Minimize	$cy$	Dual vector
	s.t.	$Ay - \alpha x = 0$	
		$By = 1$	
		$Dx = 1$	

$x, y$  are vectors of 0-1 variables

A dual of P3 can be devised by incorporating the first set of constraints into the objective function, leading to problem P4.

P4:                      Maximize               $\theta(w)$

$w$  is a vector of unrestricted variables

where                       $\theta(w) = \text{Minimum } \{(c + wA)y - w \alpha x\}$

s.t.                       $By = 1$

$Dx = 1$

$x, y$  are vectors of 0-1 variables

which could be rewritten as

$$\theta(w) = \underbrace{\left\{ \begin{array}{l} \text{Minimum } (c + wA)y \\ \text{s.t. } By = 1 \\ y \text{ is a vector of 0-1 variables} \end{array} \right\}}_I + \underbrace{\left\{ \begin{array}{l} \text{Minimum } -w \alpha x \\ Dx = 1 \\ x \text{ is a vector of 0-1 variables} \end{array} \right\}}_{II}$$

alternatively, I and II could be written in the forms:

I:    Minimum     $\sum_i \sum_j \sum_k \sum_l (c_{ijkl} + u_{ij} + v_{kl}) y_{ijkl}$

s.t.               $\sum_i \sum_k y_{ijkl} = 1 \text{ for all } j, l$

$y_{ijkl}$  is 0 or 1 for each  $i, j, k, l$

$$\begin{aligned}
\text{II. Minimum } & \sum_i -\alpha_i \sum_j (u_{ij} + v_{ij})x_{ij} \\
\text{s.t. } & \sum_i x_{ij} = 1 \quad \text{for each } j \\
& \sum_j x_{ij} = 1 \quad \text{for each } i \\
& x_{ij} \text{ is 0 or 1} \quad \text{for each } i, j
\end{aligned}$$

It is obvious that for a known vector  $w$ , whose components are the  $u_{ij}$ 's and the  $v_{kl}$ 's, part I can be evaluated by solving  $n(n-1)$  trivial knapsack problems as follows: for any given pair  $j$  and  $l$  examine  $(c_{ijkl} + u_{ij} + v_{kl})$  over all  $i$  and  $k$ . Let  $y_{ijkl} = 0$  for all except the one with the smallest  $c_{ijkl} + u_{ij} + v_{kl}$ , which has a value of 1. For part II, once  $w$  is known, its value can be obtained by solving a linear assignment problem.

The features of this formulation give rise to a simple procedure for solving the dual problem. Now we discuss the steps of this procedure.

#### Steps of the Solution Procedure, the Ascent Scheme

- 1) Choose starting values for the components of the  $w$  vector
- 2) Set  $\mu = \text{Max } \theta(w) = -\infty$ ,  $v = \text{best quadratic objective function} = +\infty$
- 3) Evaluate  $\theta(w)$  using the current  $w$  vector, update  $\mu$  as appropriate
- 4) Consider the primal solution obtained from evaluating II in Step 2, evaluate its quadratic cost. Update  $v$  as appropriate.
- 5) Compare  $\mu$  and  $v$ , if they are close enough go to Step 9, otherwise continue.
- 6) Update the components of the  $w$  vector as follows:  $w \leftarrow w + \lambda(Ay - \alpha x)$ ,

$$\text{i. e., } u_{ij} \leftarrow u_{ij} + \lambda(\sum_k \sum_l y_{ijkl} - \alpha_i x_{ij})$$

$$v_{kl} \leftarrow v_{kl} + \lambda(\sum_i \sum_j y_{ijkl} - \alpha_k x_{kl})$$

where  $\lambda > 0$  is a suitable stepsize, the choice of which will be discussed later.

- 7) Check whether the new components of the  $w$  vector are exactly the same as in the preceeding iteration, if so this means that:

$$\sum_k \sum_l y_{ijkl} - \alpha_i x_{ij} = 0 \text{ for each } i, j$$

$$\sum_i \sum_j y_{ijkl} - \alpha_k x_{kl} = 0 \text{ for each } k, l$$

i. e., both primal feasibility and dual maximum are achieved.

If this occurs, we stop and the solution in hand is the optimal solution to the problem. Otherwise continue.

- 8) Check whether a considerable number of iterations was conducted without improving  $\mu$ . If so proceed to Step 9, otherwise return to Step 3.
- 9) Terminate.

The above procedure is a subgradient optimization technique for maximizing the piecewise linear dual objective function  $\theta$ . Note that moving along  $Ay - \alpha x$  a small step will either improve  $\theta$  or at least get us closer to the optimal point. For motivation, convergence, and geometrical interpretation of general subgradient optimization procedure, the reader may refer to (Held and Karp [13], Held, Wolfe, and Crowder [14], and Bazaraa and Goode [2]).

As a result of applying this procedure either of the following three outcomes occur:

- During the solution of the dual we get to a point where primal feasibility is achieved.
- During the solution of the dual, we get a feasible primal solution, via the  $x_{ij}$ 's, whose quadratic cost is close enough to the  $\max \theta(w)$  obtained so far.
- Neither (a) nor (b) occurs and we cycle for a good number of times without a significant improvement.

If case c occurs, then we have to resort to a tree search procedure to solve the problem. However, the cost data ( $c_{ijkl}$ 's) are augmented by the values of the dual variables ( $u_{ij}$ 's and  $v_{kl}$ 's) obtained as a result of applying this procedure. This is likely to precondition the data, hence strengthen the calculated bounds and would speed up the search. In the sequel we discuss in some detail the outline of the tree search procedure.

### The Tree Search Scheme

Suppose that at any general stage of the search we have some  $x_{ij}$ 's which are fixed at value 1. This automatically fixes some values of  $y_{ijkl}$  at value 1 and some others at value 0. In particular, if  $x_{ij} = 1$ , and  $x_{kl} = 1$ , then  $y_{ijkl} = 1$  and  $y_{klij} = 1$ . We find the cost of this partial solution by calculating

$$\sum_i \sum_j \sum_k \sum_l (c_{ijkl} + u_{ij} + v_{kl}) y_{ijkl} + \sum_i \alpha_i \sum_j (u_{ij} + v_{ij}) x_{ij}$$

Now, we find a bound on the best completion of this solution, i. e.,

$$\left\{ \begin{array}{l} \text{Minimize } (c + wA)y \\ \text{s.t. } B y = 1 \\ y \text{ is a vector of 0-1} \\ \text{variables. Some } y\text{'s are} \\ \text{fixed.} \end{array} \right\} + \left\{ \begin{array}{l} \text{Minimize } -\alpha w x \\ \text{s.t. } D x = 1 \\ x \text{ is a vector of 0-1} \\ \text{variables. Some } x\text{'s are} \\ \text{fixed.} \end{array} \right\}$$

If the partial cost plus the bound on completing it is greater than or equal to the best known upper bound, the search along the current branch is terminated. Otherwise continue by trying to fix another  $x_{ij}$ . As a result of calculating the lower bound we get a feasible assignment and we calculate its quadratic cost. If the quadratic cost is less than the best known solution obtained so far, it is stored as the incumbent. Apart from these details, the general scheme of the tree search procedure would not differ much from the one given in Section IV.2.

### Choice of Initial Dual Vector $w$ and Stepsize $\lambda$

In this study an initial value  $w = 0$  was chosen. Other possibilities exist, e. g., letting  $u_{ij} = v_{ij} = -\beta_{ij}$ , where  $\beta_{ij}$  is the dual variable corresponding to the nonnegativity constraint of  $x_{ij}$  while solving the linear assignment problem with matrix  $(b_{ij})$ .

As usually is the case with subgradient optimization techniques, a small stepsize, or a decreasing sequence of stepsizes must be used. One method that has also been tried is to normalize the subgradient vector  $Ay - \alpha x$  (by dividing by its norm) and then moving a distance  $\lambda$ . A sequence of decreasing  $\lambda$ 's, say (1.0, 0.4, 0.1) may be used where we switch to a smaller value of  $\lambda$  if a certain number of trials without increasing  $\theta$  is encountered.

Since subgradient optimization methods will almost certainly encounter zigzagging (see Bazaraa and Goode [2]), taking convex combinations of the directions at two successive iterations was employed. This indeed helped reduce the zigzagging phenomena.

### Limited Experience with the Algorithm

The algorithm discussed above was tried on the 5 x 5 and 6 x 6 problems of (Nugent et. al., [21]). The 5 x 5 problem has an optimal objective of 25. The maximum value of  $\theta$  obtained is 22. In the course of dual optimization, however, the optimal solution with value 25 was obtained (Step 4). For the 6 x 6 problem whose optimal solution is 43, the ascent scheme was able to only achieve a dual objective of 36, and the optimal primal assignment with cost 43 was also obtained.

Even though the optimal solution of both problems that we tried were obtained, the lack of improvement of the dual function beyond the values of 22 and 36 was disappointing. This may be attributed to several factors.



First of all, the formulation was designed for the general nonsymmetric problem and hence did not make use of the symmetry of the problems that we tried. Secondly, the maximum of the dual objective only gives the solution of the continuous linear program of problem P2. This is due to the fact that integrality of  $x$  and  $y$  is a byproduct of the formulation after constraints (a) is placed into the objective function. Hence a duality gap is generally expected between problems P2 (or original problem) and P4. Due to the special structure of the problem, we have hoped that the duality gap (if any) is small. At this stage, we cannot confirm whether the difference between the dual objective obtained and the true optimal solution is due to a duality gap, inability to find the true maximum of the dual, or to both.

#### IV.4 A Heuristic Procedure

As will be detailed later in the computational experience section, the exact procedure was not able to find optimal solutions for large problems. A heuristic procedure is described below for obtaining "good" solutions in a reasonable computational time for larger problems.

##### Obtaining an Initial Solution

##### 1. Rank the locations as follows:

- i) Calculate  $R_i$ , the sum of distances from location  $i$  to all other locations.
- ii) Rank the locations in an ascending order of  $R_i$ .

##### 2. Rank the facilities as follows:

- i) Calculate  $E_i^1$ , the number of facilities connected to facility  $i$ .
- ii) Calculate  $E_i^2$ , the total interactions to and from facility  $i$ .
- iii) Rank the facilities in a descending order of  $E_i^1$ .

- iv) Rank the facilities in a descending order of  $E_i^2$ .
  - v) Find the combined index  $L_i$  for each facility, according to the rankings in iii) and iv).
  - vi) Rank the facilities in an ascending order of  $L_i$ .
3. At any general step  $k$ , check whether any facility remains unassigned, if so, proceed to Step 4, otherwise go to Step 7.
  4. Pick the next element in the rank, locate it to the first available location in the rank, set  $key = 1$ , and proceed to Step .
  5. Pick the element with maximum number of interactions with the most recently located facility, if none are available, go to Step 7, otherwise locate it so as to result in a minimal increase in the total cost. Set  $key = 2$  and proceed to Step 6.
  6. Exchange the element which has just been located with each of the previously located elements in turn, each time calculating the resulting change in cost. Maintain the partial assignment which causes the maximum reduction, if any; finally replace the original assignment with that one. If  $key = 1$ , return to Step 5, otherwise return to Step 3.
  7. Terminate as all elements have been assigned.

#### Improving the Initial Solution

Once a complete assignment is available, it is possible to seek improvements by interchanging the locations of two facilities. Once an improvement is achieved, it is possible to either repeat the process in search for more improvements or stop. The process can be repeated until we get to the stage where we find that no further improvements could be achieved by performing single exchanges on the pattern in hand. At each stage, it is possible to either choose the first exchange which will result in an improvement or to choose the one which yields the maximum improvements.

We have chosen the strategy of testing the effect of interchanging each facility, with each other facility, storing the best resulting pattern and interchanging the two facilities if the best pattern provides a reduction in the total cost, otherwise the same process is repeated for the next facility and so on. At any stage, if two facilities are eventually interchanged, we start all over again. However, if we fail to achieve any further improvement by interchanging the location of two facilities in the best assignment, we take the next best assignment, chosen from amongst all the assignments resulting from the test of interchanges at the immediately preceeding step, and repeat the whole procedure.

The termination criterion is simple; we can either stop when the difference between the value of the best solution so far and the lower bound is within a prespecified limit or when we have repeated the overall procedure, each time starting with a different assignment, a number of times without improvement.

#### IV.5 Computational Experience

The experience gained with the solution procedures was in relation to the eight problems of (Nugent, et. al., [21]). First, we will discuss some details pertinent to the purely computational aspects of the exact procedure.

##### a) Choice of $\alpha$

If a small value of  $\alpha$  is chosen, the fictitious upper bounds tend to be tighter and hence the search becomes faster. However, we may increase the number of restarts this way. On the other hand, if  $\alpha$  is large, then fathoming becomes weaker and the search tends to be lengthier but with a less number of restarts. The tradeoff is only computational and is data dependent.

During the course of our study, we noticed that due to the features of the search procedure, many successive good solutions are obtained very early

in the search and the optimum follows suit. As a result, we have adopted the strategy of choosing a small value of  $\alpha$  at the beginning of the search and up to a certain stage where we switch to 100% fathoming. This is conducted by specifying an initial value of  $\alpha$  and also specifying a difference between the actual upper and lower bounds at the achievement of which  $\alpha$  is switched to 100%. If the difference is appropriately chosen, we will get to the stage where the upper bound is tight enough to speed up the search and also we will not have to restart the search, once the tree is enumerated, as there is no interval of uncertainty in this case.

We found that the choice of  $\alpha = 70\%$ , as a starting value, was adequate for all the problems solved. The difference between the two bounds after which we switch to 100% fathoming varied from one problem to another.

#### b) Solving the Linear Assignment Problem

As was mentioned in IV.2, the lower bound can be calculated by different methods. One procedure involves a linear assignment problem (LAP) to obtain a tighter lower bound. There is no need, however, to solve a fresh LAP each time a lower bound is to be calculated. It is possible to take any previous solution and update it according to the new cost matrix. Consequently, each optimal solution to the LAP under a certain cost matrix could be taken as a starting feasible solution the next time a need arises to solve a LAP under a different cost matrix. Several existing codes were incorporated in a code for the exact procedure and the solution times were compared. Finally, we have selected the code of McGinnis\* to be used and the reported computational

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\*We would like to thank Mr. Leon F. McGinnis (N. C. State University) for providing us with his code which is based on the method of Glover, Klingman, and Kearny.

experience reflects its efficiency. For example, with problem 4006 (Table IV.1) the solution of the LAP was updated 26,367 times, while the total CPU time for the whole problem was about 490 seconds, i. e., under all circumstances, each updating process could not have taken more than 0.02 seconds.

### Computational Experience

In the sequel, we summarize our experience with both the exact and the heuristic procedures.

#### 1) The Exact Procedure

Tables IV.1 and IV.2 summarize the experience with the following two codes:

QAP3: a code for an algorithm based on calculating the lower bound

$C_1 + C_2 + C_3$  by matching the vectors.

QAP7: a code for an algorithm based on calculating the lower bound

$C_1 + C_2 + C_3$  by the reduction method augmented by solving a linear assignment problem only when the lower bound calculated by reduction is not higher than the current upper bound. Some of the statistics listed in these tables are:

Total Number of Nodes: is the number of nodes, both intermediate and terminal, which were generated during the search.

Total Number of Iterations: is the number of forward and backward moves conducted during the search.

Number of times it was necessary to solve an assignment problem:

whenever the lower bound calculated at any particular node by the reduction method was less than the current upper bound, it was necessary to solve a linear assignment problem in order to improve the value of this lower bound. Naturally, this strategy is applicable only to the algorithm which calculates the lower bounds by the reduction method.

Fathoming Efficiency: is the ratio between the number of times the search was not pursued, as a result of the lower bound test, to the total number of times the lower bound test was applied.

#### Comparison between QAP3 and QAP7

We recall that the only difference between QAP3 and QAP7 is the method of calculating the lower bound as shown in IV.2. From Table IV.1, we notice that the solution times when using QAP7 were always less than those obtained when QAP3 was used.

Our observation about QAP3 is that it can face severe difficulties when the problem size increases. For example, problem 4005 was rerun with a starting upper bound of 289 and 100% fathoming in hope that this will speed up the search and there will be no need to restart the search. However, we had to terminate the problem after 50,400 iterations as we noticed that 11,539 nodes were generated but the number of nodes at various levels were 0 4 41 358 2411 4101 3715 1000 184 8 3 3, i. e., a substantial part of the tree was still to be searched and the estimated time for the completion of the search was about 15 minutes on the IBM 370/165. Also, the experience with 4006 was not anymore encouraging. It may be worth mentioning, however, that to the best of our knowledge, no exact solution to problem 4005 was known before.

Note that both QAP3 and QAP7 found the optimal solution of problems 4001 through 4004 and verified optimality. QAP3 and QAP7 found the optimal of problem 4005 but only QAP7 verified optimality (this is why QAP7 used more time than QAP3 for problem 4005). It is interesting to note that the concept of stepped fathoming was essential in the procedure because otherwise the computational effort would increase significantly.

TABLE IV.1 Summary of the Computational Experience with QAP3 and QAP7

Problem Number	Size	Algorithm	Total Number of Nodes	Total Number of Iterations	No. of times the LB was calculated	No. of times it was necessary to solve an AP	Fathoming Efficiency%	Value of Best Solution Obtained	Solutions Time (seconds)*
4001	5x5	QAP3	15	38	43	--	46.12	25	0.26
		QAP7	9	20	20	14	45.00	25	0.15
4002	6x6	QAP3	27	99	129	--	54.26	43	1.01
		QAP7	18	56	67	36	52.24	43	0.63
4003	7x7	QAP3	55	177	235	--	51.06	74	3.9
		QAP7	22	62	73	40	58.90	74	2.7
4004	8x8	QAP3	231	739	1005	--	50.25	107	15.3
		QAP7	52	179	235	141	52.77	107	9.6
4005	12x12	QAP3	6877	29282**	--	--	--	289	221.8
		QAP7	5724	24496	37531	26368	50.02	289	490.4
4006	15x15	QAP3	10071	40502***	--	--	--	618	--

\*On an IBM 370/165

\*\*No restarting was allowed

\*\*\*The search was forced to termination

TABLE IV.2 Number of Nodes Generated at the Tree Levels

PROB. NUMB.	ALGORITHM	NUMBER OF ITERATIONS	TOTAL NUMBER OF NODES	LEVELS											
				1	2	3	4	5	6	7	8	9	10	11	12
4001	QAP3	38	15	0	4	5	3	3							
	QAP7	20	9	0	2	3	2	2							
4002	QAP3	99	27	0	6	13	4	2	2						
	QAP7	56	18	0	6	2	4	3	3						
4003	QAP3	177	55	0	7	21	21	3	1	2					
	QAP7	62	22	0	6	8	2	2	2	2					
4004	QAP3	739	231	0	8	40	109	58	10	3	3				
	QAP7	179	52	0	8	17	10	8	3	3	3				
4005	QAP3	29282*	6877	0	12	125	866	3367	1774	645	73	6	3	3	3
	QAP7	24496	5724	0	12	132	884	2539	1492	585	53	18	3	3	3

\*Decision tree has not been exhausted



## 2) The Heuristic Procedure

Two versions of the Heuristic Procedure were tested.

- a) QAH1: a heuristic based on the ideas mentioned in IV.4, except that it terminates once all elements have been located and the test of single exchanges is performed on the last located element only.
- b) QAH2: Multi double exchanges are performed after single exchanges cease to provide further improvements.

Computational experience with the above two heuristic procedures and their comparison with some well known heuristic methods is summarized in Table IV.3. The results of the procedures of (Hillier [15], Hillier and Connors [16], Craft [1,4], and biased sampling [21]) have been published in (Nugent, et. al., [21]), and together with the results of (Neghabat [20]) are reproduced here. QAH1 and QAH2 compare favorably with these methods, both in the quality of solution obtained and the computational time. The column designating the best solution known so far is based on either true optimal or on the best (nonreproducible) result of the biased sampling method\* of (Nugent, et. al, [21]).

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\*QAH2 obtained a solution with objective 1299 which is less than the best result known so far of 1304 for problem 4007.

TABLE IV.3 Experimental Comparisons With Some Existing Heuristic Algorithms

PROBLEM NUMBER	PROB- LEM SIZE	HILLIER		HILLIER- CONNORS		CRAFT		BIASED SAMPLING			NEG HABAT		QAH1		QAH2		BEST KNOWN COST
		AVE. COST	AVE* TIME	AVE COST	AVE* TIME	AVE COST	AVE* TIME	AVE. COST (minimum)	AVE. COST (mean)	AVE.* TIME	COST	TIME*	COST	TIME **	COST	TIME **	
4001	5	27.6	0.5	29.4	0.83	28.2	0.08	26.8	27.6	0.92	25	0.05	26	0.14	26	0.12	25***
4002	6	44.2	0.58	44.2	0.75	44.2	0.16	43.6	44.6	1.75	43	0.1	43	0.14	43	0.21	43***
4003	7	78.8	1.25	78.4	1	79.6	0.4	74.8	77.2	4.75	74	0.3	78	0.11	75	0.39	74***
4004	8	114.4	1.2	110.2	1.2	113.4	0.83	107.0	111.6	9.08	118	0.74	114	0.17	114	0.42	107***
4005	12	317.4	4.6	310.2	1.6	296.2	5.8	293.0	304.7	54.83	334	1.5	295	0.73	295	1.94	289***
4006	15	632.6	6.5	600.2	3.3	606	13.3	580.2	603.0	182.7	618	2.5	645	0.69	614	6.07	575
4007	20	1400.4	14	134.5	6.25	1339	44	1313.0	1339.0	576.3	1385	3.6	1345	1.71	1299	19.49	1304
4008	30	3267.2	33	3206.8	23.7	3189.6	262	3124.8	3189.6	3520	3260	4.5	3170	15.16	3125	71.99	3093

\*Actual or equivalent time on IBM 360/65 in seconds

\*\*Time on Univac 1108 in seconds

\*\*\*Optimal Solution

## CONCLUSION

In this section we have been concerned with solving the quadratic assignment problem. Our conclusion is that this problem is extremely difficult to solve exactly for several reasons, including the combinatorial nature of the problem, and the fact that the lower bounds calculated during a branch and bound scheme are generally very weak, leading to a lengthy search.

It was our hope to tighten the lower bounds by a duality based scheme as shown in IV.3. Even though this goal has not been achieved, we believe that further work along these lines for raising the lower bound is essential for obtaining exact and "qualified" suboptimal solutions via branch and bound.

Based on our experience, it is feasible to solve quadratic assignment problems with sizes up to  $12 \times 12$  exactly by the proposed exact procedure. However, for larger problems, it is possible to obtain good results in reasonable times by the use of the proposed heuristic (QAH2) or its curtailed version (QAH1).

## REFERENCES

1. Armour, G. C. and E. S. Buffa, "A Heuristic and Simulative Approach to Relative Location of Facilities," *Management Science*, Vol. 9, pp. 294-309, 1963.
2. Bazaraa, M. S. and J. J. Goode, "A Survey of Various Tactics for Generating Lagrangian Multipliers in the Context of Lagrangian Duality," Georgia Institute of Technology, August 1974.
3. Breuer, M. A., "The Formulation of Some Allocation and Connection Problems as Integer Problems," *Naval Research Logistics Quarterly*, Vol. 13, pp. 83-95, 1966.
4. Buffa, E. S., G. C. Armour, and T. E. Vollman, "Allocating Facilities with CRAFT," *Harvard Business Review*, Vol. 42, pp. 136-158, 1964.
5. Dorris, A. L., "The Utility of Optimization Techniques in the Design of Man-Machine Systems," Masters Thesis, Georgia Institute of Technology, 1971.
6. Francis, R. L. and J. A. White, "Facility Layout and Location: An Analytical Approach," Prentice-Hall, 1974.
7. Gaschutz, G. K. and J. H. Ahrens, "Suboptimal Algorithm for the Quadratic Assignment Problem," *Naval Research Logistics Quarterly*, Vol. 15, pp. 49-62, 1968.
8. Gavett, J. W. and N. V. Plyter, "The Optimal Assignment of Facilities to Locations by Branch and Bound," *Operations Research*, Vol. 14, pp. 210-232, 1966.
9. Gilmore, P. C., "Optimal and Suboptimal Algorithms for the Quadratic Assignment Problem," *SIAM Journal on Applied Mathematics*, Vol. 10, pp. 305-313, 1962.
10. Graves, G. W. and A. B. Whinston, "An Algorithm for the Quadratic Assignment Problem," *Management Science*, Vol. 17, pp. 453-471, 1970.
11. Hanan, M. and J. Kurtzberg, "A Review of the Placement and Quadratic Assignment Problems," *SIAM Review*, Vol. 14, pp. 324-342, 1972.
12. Heider, C. H., "An n-step, 2-variable Search Algorithm for the Component Placement," *Naval Research Logistics Quarterly*, Vol. 20, pp. 699-724, 1973.
13. Held, M. and R. M. Karp, "The Traveling Salesman Problem and Minimum Spanning Trees: Part II," *Mathematical Programming*, Vol. 1, pp. 6-25, 1971.

14. Held, M., P. Wolfe, and H. D. Crowder, "Validation of Subgradient Optimization," *Mathematical Programming*, Vol. 6, pp. 62-88, 1974.
15. Hillier, F. S., "Quantitative Tools for Plant Layout Analysis," *The Journal of Industrial Engineering*, Vol. 14, pp. 33-40, 1963.
16. Hillier, F. S. and M. M. Connors, "Quadratic Assignment Problem Algorithms and the Location of Indivisible Facilities," *Management Science*, Vol. 13, pp. 42-57, 1966.
17. Koopmans, T. C. and M. Beckman, "Assignment Problems and the Location of Economic Activities," *Econometrica*, Vol. 25, pp. 53-76, 1957.
18. Land, A. H., "A Problem of Assignment with Interrelated Costs," *Operational Research Quarterly*, Vol. 14, pp. 185-198, 1963.
19. Lawler, E. L., "The Quadratic Assignment Problem," *Management Science*, Vol. 9, pp. 586-599, 1963.
20. Neghabat, F., "An Efficient Equipment Layout Algorithm," *Operations Research*, Vol. 22, pp. 622-628, 1974.
21. Nugent, C. E., T. E. Vollman, and J. Ruml, "An Experimental Comparison of Techniques for the Assignment of Facilities to Locations," *Operations Research*, Vol. 16, pp. 150-173, 1968.
22. Pegels, C. C., "Plant Layout and Discrete Optimizing," *International Journal of Production Research*, Vol. 5, pp. 81-92, 1966.
23. Pierce, J. F. and W. B. Crowston, "Tree Search Algorithms for Quadratic Assignment Problems," *Naval Research Logistics Quarterly*, Vol. 18, pp. 1-36, 1971.
24. Steinberg, L., "The Backboard Wiring Problem: A Placement Algorithm," *SIAM Applied Mathematics*, Vol. 3, pp. 37-50, 1961.
25. Vollman, T. E., C. E. Nugent, and R. L. Zartler, "A Computerized Model for Office Layout," *The Journal of Industrial Engineering*, Vol. 19, pp. 321-329, 1968.
26. Whitehead, B. and M. Z. Elders, "An Approach to the Optimum Layout of Single Story Buildings," *Architect's Journal*, Vol. 139, pp. 1373-1380, 1964.

## V. THE TRAVELING SALESMAN PROBLEM

As discussed in Section II, the traveling salesman problem is a special important case of the quadratic assignment problem. In this section, the dual of the problem is formulated and then solved by a subgradient optimization technique. In the presence of a duality gap, a branch and bound scheme is devised to find optimal and suboptimal tours.

### V.1 Formulation of the Problem and its Dual

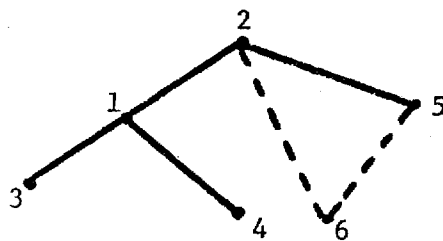
Suppose that a salesman is given a list of  $n$  cities, where each pair of cities,  $i$  and  $j$ , is connected by a link of length  $c_{ij}$  (if cities  $i$  and  $j$  are not directly linked, let  $c_{ij} = +\infty$ ). The problem is then to find a tour of minimal length which the salesman can follow so as to visit each city exactly once, and then return to where he started.

Let  $x_{ij}$  be a variable associated with link  $ij$ .  $x_{ij}$  is equal to 1 if the salesman travels from city  $i$  to city  $j$  and is 0 otherwise. The problem can be stated as follows:

$$\begin{array}{ll}
 \text{Minimize} & \sum_{\substack{j=1 \\ j \neq i}}^n \sum_{i=1}^n c_{ij} x_{ij} \\
 \text{Subject to} & \sum_{\substack{j=1 \\ j \neq i}}^n x_{ij} = 1 \quad i = 1, 2, 3, \dots, n & \text{V.1} \\
 & \sum_{\substack{i=1 \\ i \neq j}}^n x_{ij} = 1 \quad j = 1, 2, 3, \dots, n & \text{V.2} \\
 & x_{ij} \text{ is 0 or 1} \quad i = 1, \dots, n; j = 1, \dots, n & \text{V.3} \\
 & \text{No subtours}
 \end{array}$$

Since the restrictions V.1, V.2, and V.3, by themselves, may allow the meaningless answer of having two or more disjoint subtours, a restriction eliminating subtours must be explicitly introduced. There are various methods in the literature for enforcing this restriction. We will adopt the 1-tree method of (Held and Karp [7]).

Let us begin by reviewing the definition of a 1-tree [7]. A 1-tree of the graph composed of the  $n$  cities and their interconnecting links is a tree which spans  $n - 1$  cities, together with two distinct links from the remaining city, to any two of the  $n - 1$  cities spanned by the tree. Therefore, by definition, a 1-tree contains exactly one cycle. Of course, the number of links in the cycle is less than or equal to  $n$ . A 1-tree then corresponds to a tour if and only if the cycle formed by the 1-tree has  $n$  links. A 1-tree which is composed of a tree covering nodes 1, 2, 3, 4, and 5 plus two links from node 6 to the tree is shown below.



An Example of a 1-Tree

Now note that if constraints V.1, V.2, and V.3 are satisfied then there is at least one cycle. If there is only one cycle then it is a tour. By definition of the 1-tree, it admits exactly one cycle. Therefore, constraints V.1, V.2, and V.3, coupled with the constraint that the vector  $x$  forms a 1-tree, admit no subtours. Therefore, the traveling salesman problem can be formulated as follows:

$$\begin{aligned}
&\text{Minimize} && \sum_{j=1}^n \sum_{\substack{i=1 \\ j \neq i}}^n c_{ij} x_{ij} \\
&\text{Subject to:} && \sum_{\substack{j=1 \\ j \neq i}}^n x_{ij} = 1 && i = 1, 2, \dots, n \\
&&& \sum_{\substack{i=1 \\ i \neq j}}^n x_{ij} = 1 && j = 1, 2, \dots, n \\
&&& x \in X
\end{aligned}$$

where  $X$  is the set of 1-trees spanning the graph, i. e.,

$X = \{(x_{11}, \dots, x_{1n}, \dots, x_{nn}) : x_{ij} = 0 \text{ or } x_{ij} = 1 \text{ (} i \neq j \text{), and}$   
the links  $ij$  with  $x_{ij} = 1$  form a 1-tree}.

Note that there are very efficient procedures for finding a 1-tree with minimal cost [5,11]. This will involve pulling one node out of the  $n$  nodes, which we will refer to as the initial node, and then:

1. finding a tree of minimal length which spans the remaining  $n - 1$  nodes
2. finding two links of minimal length which connect the initial node to the spanning tree.

Any node can be used as the initial node, and we will discuss later a procedure for making a "good" choice of the initial node. Calculating a 1-tree with minimal cost, in the presence of restrictions V.1, and V.2, namely the traveling salesman problem, is indeed a difficult problem. If we handle the restrictions V.1, and V.2 via lagrangian multipliers (dual variables) and put them into the objective function, we obtain the following dual formulation.



### Dual Problem

Maximize  $\theta(u,v)$

$u$  and  $v$  are unrestricted

where,

$$\begin{aligned}\theta(u,v) &= \text{Minimum}_{x \in X} \left\{ \sum_j \sum_i c_{ij} x_{ij} + \sum_i u_i \left( \sum_j x_{ij} - 1 \right) + \sum_j v_j \left( \sum_i x_{ij} - 1 \right) \right\} \\ &= \text{Minimum}_{x \in X} \left\{ \sum_j \sum_i (c_{ij} + u_i + v_j) x_{ij} - \sum_i u_i - \sum_j v_j \right\} \\ &= -\sum_i u_i - \sum_j v_j + \text{Minimum}_{x \in X} \sum_j \sum_i (c_{ij} + u_i + v_j) x_{ij}\end{aligned}$$

Note that evaluating  $\theta(u,v)$ , for fixed values of  $u$  and  $v$  is an easy task since it involves finding a minimal 1-tree, where the cost  $c_{ij}$  of the link  $ij$  is replaced by  $c_{ij} + u_i + v_j$ . Also note that in finding the minimal 1-tree, we are not concerned about the orientation of the links. In other words, we are actually searching for a minimal-tree, rather than an arborescence. This simplifies the problem of evaluating  $\theta(u,v)$  considerably. Of course, we may think of the vectors  $u$  and  $v$  as penalties for violating the proper orientation of the links. So if when calculating the minimal 1-tree, too many links "left" node  $i$  we will attempt to "penalize" this deficiency in the next iteration by increasing  $u_i$ . Similarly, if too many links "entered" node  $j$ , then we will attempt to "penalize" this deficiency by increasing  $v_j$ .

### Simplification of the Symmetric Case

In the symmetric case, as long as the vector  $x$  forms a 1-tree, and since the orientation of links is not important, we no longer require that the number of links leaving node  $i$  is 1, and the number of links entering node  $i$  is 1, as long as the number of links incident to node  $i$  is equal to 2. The primal problem can thus be reformulated as follows:

$$\begin{aligned}
& \text{Minimize} && \sum_{j=1}^n \sum_{\substack{i=1 \\ j \neq i}}^n c_{ij} x_{ij} \\
& \text{Subject to} && \sum_{j=1}^n \sum_{\substack{i=1 \\ j \neq i}} x_{ij} + \sum_{j=1}^n \sum_{\substack{i=1 \\ j \neq i}} x_{ji} = 2 \quad \text{for } i = 1, 2, \dots, n \\
& && x \in X
\end{aligned}$$

where  $X$  is the collection of 1-trees. The above restrictions ensure forming a tour. If the tour is not "oriented," it can be made so, by simply reorienting some of the links, without affecting the cost. The dual of the above problem is given below, with only one set of dual variables given by the vector  $u$ .

$$\text{Maximize} \quad \theta(u)$$

$u$  is unrestricted

where,

$$\begin{aligned}
\theta(u) &= \text{Minimum}_{x \in X} \sum_{j=1}^n \sum_{\substack{i=1 \\ j \neq i}}^n c_{ij} x_{ij} + \sum_{i=1}^n u_i \left( \sum_{\substack{j=1 \\ j \neq i}}^n x_{ij} + \sum_{\substack{j=1 \\ j \neq i}}^n x_{ji} - 2 \right) \\
&= -2 \sum_{i=1}^n u_i + \text{Minimum}_{x \in X} \sum_{j=1}^n \sum_{\substack{i=1 \\ j \neq i}}^n (c_{ij} + u_i + u_j) x_{ij}
\end{aligned}$$

From now on we will concentrate on the nonsymmetric case with dual variables  $u$  and  $v$ . For the symmetric case, the dual problem can be simplified as illustrated above, with no need for the dual vector  $v$ .

## V.2 Maximization of $\theta$

Let the collection of 1-trees with a fixed initial node be  $x_1, x_2, \dots, x_t$ . Then  $\theta(u, v)$  can be written as follows:

$$\begin{aligned}
\theta(u, v) &= \text{Minimum}_{x \in X} \sum_i \sum_j c_{ij} x_{ij} + \sum_i u_i \left( \sum_j x_{ij} - 1 \right) + \sum_j v_j \left( \sum_i x_{ij} - 1 \right) \\
&= \text{Minimum}_{x \in X} cx + ug(x) + vh(x) \\
&= \text{Minimum}_{1 \leq j \leq t} cx_j + ug(x_j) + vh(x_j)
\end{aligned}$$

where  $g_i(x) = \sum_j x_{ij} - 1$  and  $h_i(x) = \sum_j x_{ji} - 1$  for  $i = 1, 2, \dots, n$ . Note that  $\theta$  is the minimum of a finite number of affine functions and is hence a piecewise linear concave function which is finite everywhere.

Several procedures for maximizing  $\theta$  are available. Here a subgradient optimization procedure similar to that of (Held and Karp [8]) is used. Given a vector  $(u, v)$ , a minimal 1-tree  $x$  is found, where the cost of link  $ij$  is  $c_{ij} + u_i + v_j$ .  $(u, v)$  is updated by moving a discrete step along the subgradient  $\{g(x), h(x)\}$  and the process is repeated. Motivation, convergence, and geometrical aspects of the procedure can be found in [8], [9], and [1].

The initial node used throughout the algorithm is determined as follows.  $n$  minimal 1-tree problems with different initial nodes are solved with  $u = v = 0$ . The initial node whose minimal 1-tree has maximal objective is used. The following is a precise statement of the ascent algorithm used for maximizing the dual function (with obvious modification for the symmetric problem).

### Subgradient Optimization of the Dual

#### Initialization Step

Choose  $u_1 = v_1 = 0$ . Choose a sequence of stepsizes  $(\lambda_1, \lambda_2, \dots, \lambda_t)$  where  $\lambda_{j+1} < \lambda_j$  for  $j = 1, 2, \dots, t-1$ . Choose the maximum number of iterations  $v_{\max}$  allowed without improving the incumbent  $\theta^*$ . Let  $j = k = 1$ ,  $v = 0$ ,  $u^* = u_1$ ,  $v^* = v_1$ , and  $\theta^* = -\infty$ . Go to Step 1.

#### Step 1 Finding $\theta(u_k, v_k)$

Solve the following minimal 1-tree subproblem.

$$\text{Minimize} \quad \sum_{\substack{j=1 \\ j \neq i}}^n \sum_{i=1}^n (c_{ij} + u_{ik} + v_{jk}) x_{ij}$$

Subject to:  $x \in X$

where  $u_k = (u_{1k}, \dots, u_{nk})$  and  $v_k = (v_{1k}, \dots, v_{nk})$ .

Let  $x_k$  be an optimal 1-tree. If  $x_k$  is a directed tour, then stop, the optimal primal and dual solutions have been found. Otherwise,

$$\text{let } \theta(u_k, v_k) = \sum_{j=1}^n \sum_{i=1}^n (c_{ij} + u_{ik} + v_{jk}) x_{ij} - \sum_{i=1}^n u_{ik} - \sum_{j=1}^n v_{jk}.$$

If  $\theta(u_k, v_k) > \theta^*$ , then let  $v = 0$ , replace  $\theta^*$  by  $\theta(u_k, v_k)$ , replace  $(u^*, v^*)$  by  $(u_k, v_k)$ ,  $x^*$  by  $x_k$ , and go to Step 2. If on the other hand,  $\theta(u_k, v_k) \leq \theta^*$ , replace  $v$  by  $v+1$ . If  $v = v_{\max}$ , go to Step 3. If  $v < v_{\max}$ , go to Step 2.

#### Step 2 (Update dual variables)

Let  $u_{k+1} = u_k + \lambda_j g(x_k)$  and  $v_{k+1} = v_k + \lambda_j h(x_k)$ . Replace  $k$  by  $k+1$  and repeat Step 1.

#### Step 3 (reduce stepsize)

If  $j = t$  then stop with  $\theta^*$  and  $(u^*, v^*)$ . Otherwise replace  $j$  by  $j+1$  and  $v$  by 0. Let  $u_{k+1} = u^* + \lambda_j g(x^*)$  and  $v_{k+1} = v^* + \lambda_j h(x^*)$ . Replace  $k$  by  $k+1$  and repeat Step 1.

### V.3 The Branch and Bound Scheme

So far we have described a procedure for finding the optimal (or near optimal) dual solution. In the process, an optimal tour for the traveling salesman problem may be found. If this were not the case, a branch and bound scheme for finding an optimal directed tour will be described. It is important to note that even in this case, optimizing the dual function played the following two significant roles:

1. Provided a "tight" lower bound on the optimal traveling salesman problem and hence making the notion of "qualified" suboptimal solutions attractive.
2. Rather than using the cost  $c_{ij}$  of link  $ij$  in the branch and bound scheme,  $c_{ij}$  is replaced by  $c_{ij} + u_i + v_j$ , where  $u$  and  $v$  are the

optimal dual variables, thus resulting in a "well-conditioned" cost matrix. The computational effort for finding the optimal tour, using the new cost matrix  $c_{ij} + u_i + v_j$  is considerably smaller than the effort to be expended, had the original cost matrix been used in a branch and bound scheme. This is due to the fact that the new cost matrix  $(c_{ij} + u_i + v_j)$  is "conditioned" in the sense that an optimal 1-tree with the above matrix is "close" to the optimal traveling salesman tour.

Consider now the following traveling salesman problem:

$$\begin{aligned}
 &\text{Minimize} && \sum_{\substack{j=1 \\ j \neq i}}^n \sum_{i=1}^n (c_{ij} + u_i + v_j) x_{ij} \\
 &\text{Subject to:} && \sum_{\substack{j=1 \\ j \neq i}}^n x_{ij} = 1 && \text{for } i = 1, 2, \dots, n \\
 &&& \sum_{\substack{i=1 \\ i \neq j}}^n x_{ij} = 1 && \text{for } j = 1, 2, \dots, n \\
 &&& \text{No Subtours}
 \end{aligned}$$

Note that an optimal tour obtained by the above problem is also an optimal tour of the original traveling salesman problem.

#### General Framework

A list of distinct nodes  $i_1, i_2, i_3, \dots, i_k$  and corresponding links  $(i_1, i_2), \dots, (i_{k-1}, i_k)$  form a partial solution. Since the nodes are distinct, no subtours are a part of the partial solution. If  $k = n$ , then a complete tour is obtained by adding the link  $(i_k, i_1)$ . Suppose that we are given a partial solution, and attempt to find a feasible completion. A lower bound on the cost of completing this partial solution is computed, say  $B$ . If  $B \geq C^*$ , where  $C^*$  is the cost of the best tour known so far, then the partial solution is fathomed. If  $B < C^*$  then an additional link  $(i_k, i_{k+1})$  is added to the partial tour where  $i_{k+1}$  is not equal to  $i_j$  for  $j = 1, 2, \dots, k$ .

### Calculation of the Lower Bound

Suppose that  $i_1, i_2, \dots, i_k$  form a partial solution. If  $(i_j, i_{j+1})$  for  $j = 1, 2, \dots, k-1$  are to be a part of a tour, then the remaining nodes must themselves form a "string", which is connected to nodes  $i_1$  and  $i_k$ . Since the "string" is a tree spanning the remaining nodes, then its cost must be greater or equal to the minimal cost of all trees spanning these nodes. Also, the cost of connecting the partial solution to this open loop is greater or equal to the sum of the following:

1. Minimal cost of (available) links leaving node  $i_k$
2. Minimal cost of (available) links entering node  $i_1$

We use the word available, since links  $(i_j, i_1)$  for  $j = 2, \dots, k$  and  $(i_k, i_j)$  for  $j = 1, 2, \dots, k-1$  are not available, because their use will form a subtour. Also, as the search continues, some other links will be prohibited from usage. From this it is clear that a lower bound on the cost of the partial solution, plus the cost of its completion, is given by:

$$B = C_1 + C_2 + C_3, \text{ where}$$

$C_1$  : the actual cost of the partial solution

$$C_1 = \sum_{j=1}^{k-1} (c_{i_j i_{j+1}} + u_{i_j} + v_{i_{j+1}})$$

$C_2$  : the cost of the minimal tree spanning the remaining nodes, where the cost of link  $ij$  is  $c_{ij} + u_i + v_j$ .

$C_3$  : Minimal cost of "connecting" the partial solution to the spanning tree.

$$C_3 = \text{Minimum}_{j \notin A} (c_{i_k j} + u_{i_k} + v_j) + \text{Minimum}_{j \notin B} (c_{j i_1} + u_j + v_{i_1})$$

where  $A$  is the set of nodes which  $i_k$  is not eligible to enter, and

$B$  is the set of nodes which are not eligible to enter node  $i_1$ .

An ascent scheme could have been used to calculate the lower bound as in [8]. This is not done here, however, since the lower bound obtained above is easy to obtain and seems to be tight enough.

### Continuation of the Search

#### 1. Fathoming (Backward Move)

Suppose that we have a partial tour  $i_1, i_2, \dots, i_k$ ; i. e.,  $x_{i_1 i_2} = x_{i_2 i_3} = \dots = x_{i_{k-1} i_k} = 1$ . The level of the search is called  $k - 1$ . The lower bound  $B$  on the cost of the partial solution plus its completion is calculated as shown above. If  $B \geq C^*$ , where  $C^*$  is the current best known cost of a complete tour, then the partial solution is fathomed. The last move from  $i_{k-1}$  to  $i_k$  is prohibited, i. e.,  $x_{i_{k-1} i_k} = 0$ . A new bound  $B'$  on completing the cost of the partial tour  $i_1, \dots, i_{k-1}$  with the additional restriction that  $x_{i_{k-1} i_k} = 0$  is computed. This is done in the same manner as before, except of course that  $i_k$  now belongs to the remaining nodes and the link  $i_{k-1} i_k$  cannot be used. If  $B' \geq C^*$ , then the partial solution  $i_1, i_2, \dots, i_{k-1}$ , with the added restriction  $x_{i_{k-1} i_k} = 0$  can lead to no improved solutions. This means that all possibilities at level  $k - 1$  are exhausted. This case will be called a strong fathoming, and the level of the tree is reduced by 1 unit. In this case  $x_{i_{k-1} i_k} = 0$  is not prohibited any more, whereas  $x_{i_{k-2} i_{k-1}}$  is forced to be zero. The process is then repeated. If on the other hand  $B' < C^*$ , which will be called weak fathoming, then we seek to find a node  $j$  such that  $x_{i_{k-1} j} = 1$ . This step will be discussed in more detail in the forward move below.

## 2. Forward Move

If the lower bound is less than  $C^*$ , then it may be possible to find a completion of the partial solution with an objective less than  $C^*$ . Suppose that the current level of the tree is  $k - 1$ , where at level  $k - 1$  we have  $x_{i_{k-1}i_k} = 1$ . Now we seek a node  $i_{k+1} \neq i_j$  for  $j = 1, 2, \dots, k-1$  and the move  $i_k i_{k+1}$  is not prohibited from previous fathoming (if such a node  $i_{k+1}$  is not found then the partial solution is fathomed and the level of the tree is reduced). This rule is used: choose  $i_{k+1}$  which minimizes  $c_{i_k j} + u_{i_k} + v_j$  for all available  $j$ . The level of the tree is increased by 1, a lower bound is calculated, and the process is repeated. Needless to say, if the level is  $n - 1$ , then  $i_n$  is joined with  $i_1$  and the cost of the link  $i_n i_1$  is added to the cost of the partial tour. If the total cost is less than  $C^*$ , then  $C^*$  is replaced by the total cost, and the tour is stored as the best tour obtained so far.

## 3. Termination

We have described forward and backward progress of the search. If the level of the search ever reaches value zero, then we stop. This means that all possible completions of  $x_{i_1 i_2} = 1$  and  $x_{i_1 i_2} = 0$  have been enumerated, i. e., all the search space has been exhausted. The stored tour and  $C^*$  give the optimal solution.

### Summary of the Algorithm

#### Initialization Step

Let  $P(i) = \emptyset$  for all  $i = 1, 2, \dots, n$  and let  $C^* = \infty$ . Choose node  $i_1$  (according to the procedure described earlier) as the initial node in the maximization of  $\theta$ . Let  $k = 0$  and go to Step 1.



### Step 1 (Forward Move)

If  $k = n - 1$ , replace  $C_1$  by  $C_1 + (c_{i_n i_1} + u_{i_n} + v_{i_1})$ . If  $C_1 < C^*$ , then replace  $C^*$  by  $C_1$ , store the complete solution  $i_1, \dots, i_n, i_1$  and go to Step 2. If  $C_1 \geq C^*$ , go to Step 2. If  $k < n - 1$ , calculate a lower bound  $B$  on all completions of the partial solution at hand, where  $B = C_1 + C_2 + C_3$ ,  $C_1$ ,  $C_2$ , and  $C_3$  are calculated as described above,  $A = P(i_k) \cup \{i_1, \dots, i_{k-1}, i_k\}$  and  $B = \{i_1, i_2, \dots, i_k\}$ . If  $B \geq C^*$ , go to Step 2. Otherwise, pick  $i_{k+1} \notin A$  such that  $c_{i_k i_{k+1}} + u_{i_k} + v_{i_{k+1}} = \text{Minimum}_{j \notin A} (c_{i_k j} + u_{i_k} + v_j)$ . Replace  $k$  by  $k + 1$  and repeat Step 1.

### Step 2 (Fathom)

Replace  $P(i_{k-1})$  by  $P(i_{k-1}) \cup \{i_k\}$ . Calculate a lower bound  $B$  on all completions of the partial solution  $i_1, i_2, \dots, i_{k-2}, i_{k-1}$ . If  $B \geq C^*$ , go to Step 3. Otherwise, choose a node  $j \notin A$  such that  $c_{i_{k-1} j} + u_{i_{k-1}} + v_j = \text{Minimum} \{c_{i_k i} + u_{i_{k-1}} + v_i : i \notin A\}$ . This forms a new partial solution, where  $i_k$  is replaced by  $j$ . Go to Step 1.

### Step 3 (Strong Fathoming)

$i_{k-1}$  is placed in  $P(i_{k-2})$ .  $P(i_{k-1})$  is replaced by  $\emptyset$ , and  $k$  is replaced by  $k - 1$ . If  $k = 0$ , go to Step 4, otherwise go to Step 2.

### Step 4 (Termination)

Search has been completed. Optimal cost is  $C^*$  and its corresponding tour is  $i_1, i_2, \dots, i_n, i_1$ . Stop.

## V.4 Stepped Fathoming

Rather than attempting to fathom a partial solution when the lower bound  $B$  on its completion is greater than or equal to the best known  $C^*$ , we may fathom if  $B$  is "close" to  $C^*$ . This will enable us to fathom quickly

and hence reduce the portion of the search tree explicitly enumerated. This procedure can be controlled either to obtain suboptimal solutions, with any desired degree of accuracy or to obtain optimal solutions. In both cases the computational effort will be significantly reduced. The following two methods are proposed for implementing this notion.

#### Method 1

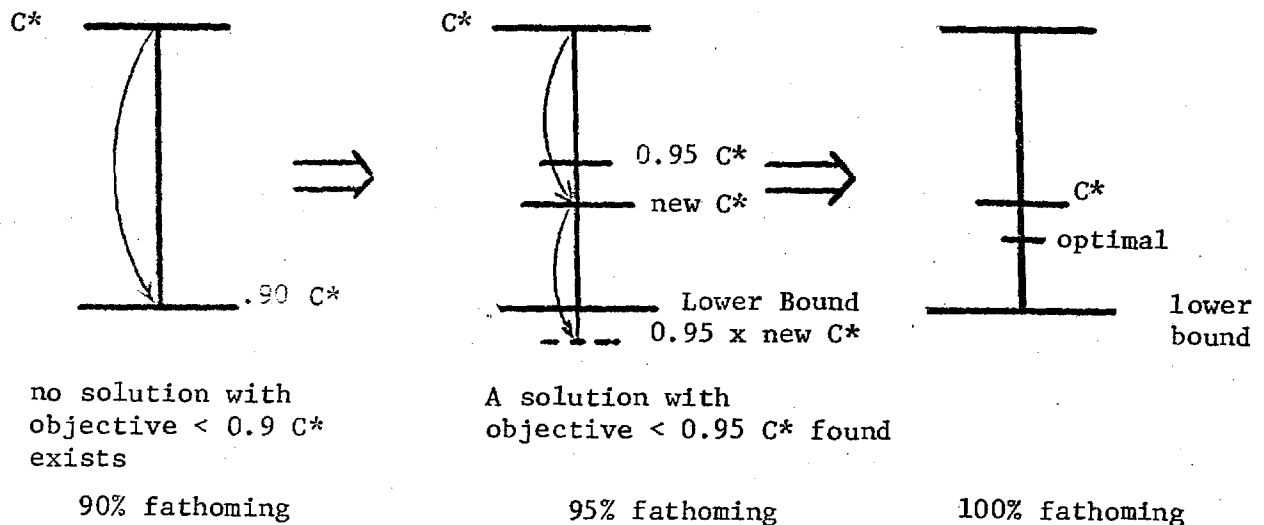
Suppose that a partial solution is fathomed if  $B \geq \alpha C^*$ , where  $\alpha \in (0,1]$ . In this case, the partial solution is abandoned if there is no hope that it leads to a cost which is better than  $\alpha C^*$ . The objective of this simple strategy is clear. We want to fathom the partial solution quickly, even if it may lead to a slight improvement. As the reported computational experience shows, this simple strategy speeds the computational effort considerably. Of course, as a new  $C^*$  is found, then we fathom whenever the bound is greater or equal to  $\alpha$  times the new  $C^*$ . The procedure continues until we cannot find a feasible solution with an objective less than  $\alpha C^*$ . So we have a feasible tour with objective  $C^*$ , coupled with the statement that the optimal objective is greater than or equal to  $\alpha C^*$ . Needless to say, if  $\alpha C^*$  is less than the overall lower bound, which is  $\text{Max } \theta(u,v)$ , then there is no need to complete the search, since we know that there are no solutions with objective less than  $\alpha C^*$ .

#### Choice of $\alpha$

If  $\alpha$  is small, fathoming will speed up considerably resulting in a small computational effort. But on the other hand, the quality of the best obtained feasible solution is not guaranteed. A tradeoff between the desired solution quality and the computational effort expended exists. We recommend values of  $\alpha \geq 0.93$  depending on the accuracy required.

### Exact Solution by Stepped Fathoming

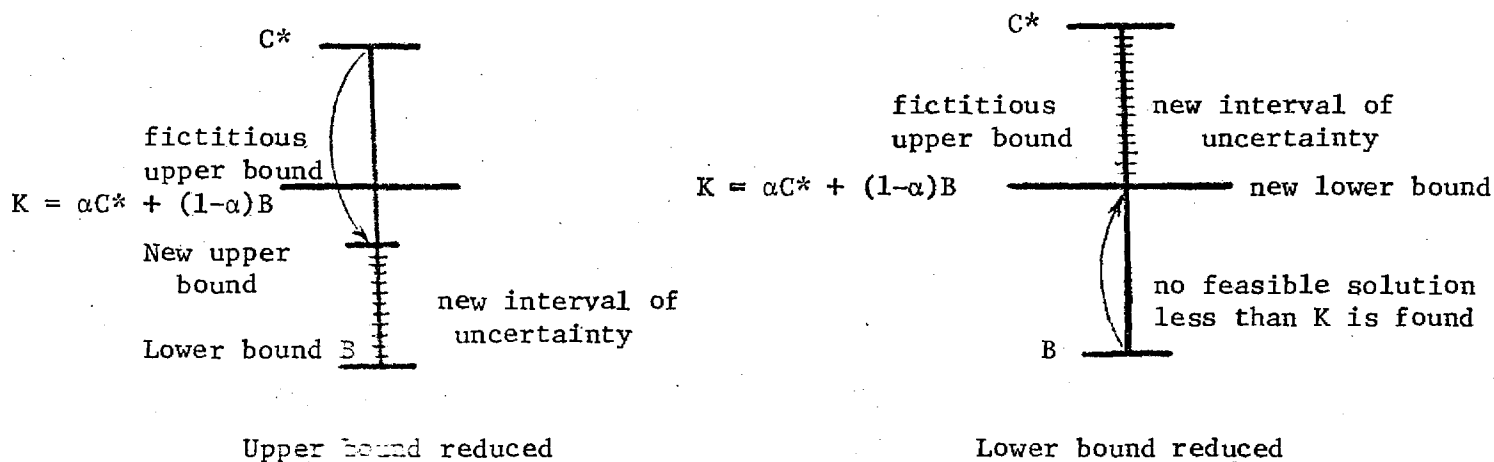
A slight modification of the above scheme can be used to find the exact optimal solution. Suppose that an initial  $\alpha_1$  is used, say  $\alpha_1 = 0.90$ . Eventually we achieve a final  $C^*$  and cannot find a feasible solution with objective less than  $0.90 C^*$ , i. e.,  $0.9 C^*$  is a lower bound on all solutions. We either stop here, or else switch to  $\alpha_2 \geq \alpha_1$ , say  $\alpha_2 = 0.95$ . We start the search from the stored best solution corresponding to  $C^*$ , which also includes all the prohibited visitations. We are either able to find a feasible solution with objective less than  $0.95 C^*$  (none less than  $0.9 C^*$  exist), and repeat the process, or else conclude that none exist. We may then switch to 97% fathoming and eventually 100% fathoming. This process is illustrated below and proved very efficient from a computational point of view.



### Method 2

At each stage of the algorithm, we have an upper bound  $C^*$ . A lower bound on the overall problem, namely  $\text{Max } \theta(u,v)$  is also known. Suppose we fathom on  $K = \alpha C^* + (1 - \alpha)B$ , where  $\alpha \in (0,1]$ . Two cases are possible. In the first case, we will be able to find a tour with objective less than  $K$ . The objective of this new solution becomes the new upper bound

$C^*$  and the process is repeated. In the second case, we will not be able to find such a solution. This automatically implies that there are no solutions with objective less than  $K$ , and hence  $K$  itself is the new lower bound. The process is then repeated. The gap between the lower and upper bounds is continuously reduced either by lowering the upper bound or by raising the lower bound. When the difference between them is smaller than a prescribed tolerance, we either stop, or try to find solutions with objective less than  $C^*$  itself, i. e., we switch to  $\alpha = 1$ . In the former case we obtain a "qualified" suboptimal solution, and in the latter case we obtain the optimal solution.



#### Choice of $\alpha$

$\alpha$  is any number in the interval  $(0,1]$ . Of course, if  $\alpha$  is close to 1, then we are in effect fathoming on a number very close to  $C^*$ , and the search will not speed considerably. On the other hand, if  $\alpha$  is close to zero, then we only fathom if we obtain a feasible solution very close to the overall lower bound. In this case, fathoming will be fast, but it is likely not to obtain feasible solutions less than  $K$ . The value  $\alpha = 0.5$  is recommended so that the distance of uncertainty is halved at each time.

Computational effectiveness of stepped fathoming will be clear in the next section. For more details on the application of stepped fathoming in tree search algorithms the reader may refer to [2].

#### V.5 Computational Experience

This section summarizes the computational experience with the algorithm discussed in V.3 for both symmetric and nonsymmetric problems. We attempted to solve some standard problems in the literature as well as some problems generated at random. The following three standard symmetric problems were solved:

1. 25 city problem of (Held and Karp [6]).
2. 42 city problem of (Dantzig, Johnson, and Fulkerson [4]).
3. 57 city problem of (Karg and Thompson [10]).

In addition, several problems ranging from 30 to 80 cities have been generated at random from a uniform distribution with lower and upper limits of 100 and 3400. All computations were executed on a Univac 1108 machine and the reported times are in seconds.

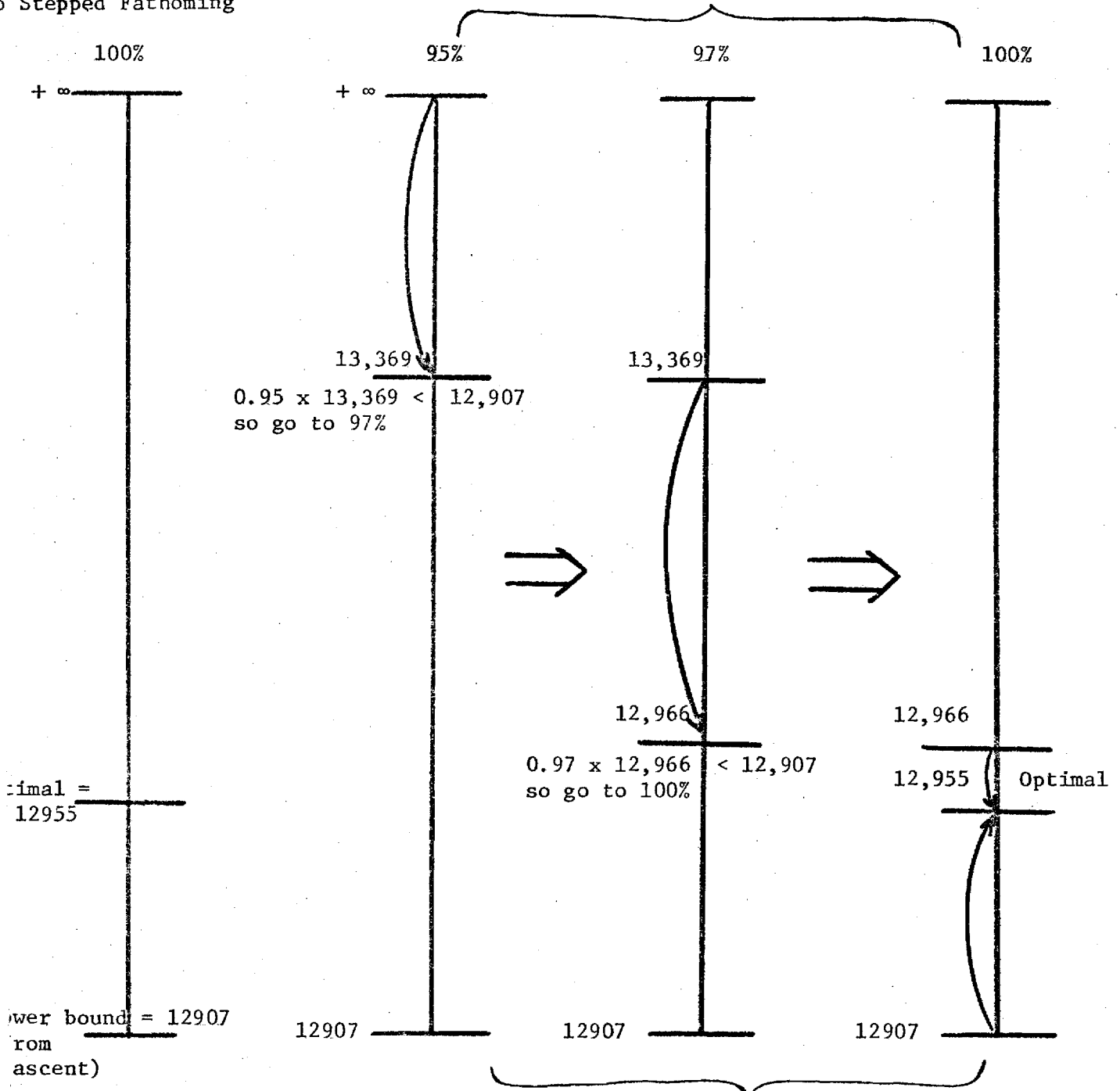
During the course of dual optimization (or ascent) two sequences of stepsizes have been used, namely (5.0, 1.0, 0.2) and (5.0, 1.0). The stepsize is reduced if 20 trials (in some instances 30 trials were used) to improve the best known dual objective failed to do so. The sequence (5.0, 1.0, 0.2) with 30 failures managed to attain a higher dual value, of course, with more computational effort spent on the ascent. This usually resulted in less time expended in the branch and bound phase of the algorithm, however. From our experience it is very difficult to ascertain whether it is advantageous to have the smallest stepsize be 1.0, 0.2, or even less than 0.2. It is safe to say, however, that the extra effort used to reduce the stepsize to 0.2 is not lost as far as the overall computational effort is concerned.

In the branch and bound phase, the best multipliers of the ascent are used to obtain a modified cost matrix. As described in the last section, no ascent is used to calculate the bounds in this phase (as opposed to Held and Karp [8]). This, of course, increases the portion of the tree explicitly searched, but reduces the effort in calculating the bounds of completing the partial solutions. Also, the scheme of stepped fathoming was implemented and found very effective in reducing the portion of the tree to be enumerated.

We will use the 57 city problem to illustrate the concept of stepped fathoming. The results with and without employing the scheme are shown below. Note that the best dual objective of 12907 was obtained on ascent in 17 seconds. Using 100% fathoming (i. e., without stepped fathoming), the optimal tour with cost 12955 was obtained in an additional 708 seconds. The same optimal was obtained (and verified) using a stepped fathoming of 95%, 97%, and then 100% in an additional 60 seconds resulting in a total computational effort of 77 seconds.

# Stepped Fathoming

## Stepped Fathoming



Total Computational Time = 725 seconds

Total Computational Time = 77.2 seconds

### Results of the 25, 42, and 57 City Problems

Table V.1 summarizes the results of the 25, 42, and 57 city problems. The last column gives the computational time reported by (Held and Karp [8]) to obtain the optimal solutions (time in seconds on an IBM 360/91). Stepped fathoming with 95%, 97%, and 100% is used here to find and verify the optimal solutions. Note that in the 25 city problem no duality gap exists. A branch and bound phase was used, however, since an alternative optimal 1-tree which is not a tour was found on ascent.

The branch and bound procedure can be used to obtain suboptimal solutions with any prescribed apriori degree of accuracy. The computational effort will usually increase as a function of the accuracy desired. To illustrate, a 95% fathoming was used to solve the 42 city problem. A tour with cost 708 is obtained in 6 seconds. The percentage deviation of the optimal = 699 from the 95% solution 708 is 1.2% (less than the theoretical bound of 5%). Even if the true optimal solution were not known, then in 6 seconds we would have obtained a tour with cost 708 coupled with the statement that the optimal solution is greater than or equal to 694.

It is interesting to note, however, that  $\alpha_1 > \alpha_2$  does not necessarily imply that quality of an  $\alpha_1\%$  solution will be better than an  $\alpha_2\%$  solution or that the computational effort of an  $\alpha_1\%$  solution will be larger than an  $\alpha_2\%$  solution. This is due to the fact that the quality of the solution and the computational effort depend on many other factors such as the lower bound from the ascent phase, the new cost matrix, the intermediate feasible solutions obtained, etc.

### Symmetric Random Problems

The two methods of stepped fathoming discussed in V.4 were used to obtain suboptimal solutions of randomly generated problems of sizes ranging



TABLE V.1

Problem Size	Dual objective before ascent	Dual objective after ascent	Time for ascent (seconds)	Optimal primal objective	Time for branch and bound (seconds)	Total Computational time (seconds)	Computational time reported by Held and Karp
25	1453	1711	1.4	1711	0.2	1.6	12
42	629	694	5.3	699	22.5	27.8	54
57	11520	12907	17.2	12955	60.0	77.2	780

from 40 to 80 cities. In many cases, optimality is verified during the course of optimization since the primal solution equals the lower bound obtained from ascent. These are denoted by \*.

#### Method 1

Stepped fathoming with 95%, 97%, and then 99% was used on problems of size 40, 50, and 60. On problems of size 70 a stepped fathoming of 95% and then 98% is used. Finally, a 95% fathoming is used in case of the 80 cities. We chose not to obtain 100% solutions because the computational effort increases significantly with very little improvement (or none at all) in the quality of solutions. The results of method 1 are summarized in Table V.2.

#### Method 2

The same random problems shown above were solved by method 2 where the interval of uncertainty is halved at each stage until the difference between the lower bound and the upper bound (best known feasible solution) is less than or equal to 100. The quality of solutions and the computational effort is comparable to that of method 1. A summary of the results is shown in Table V.3.

#### Nonsymmetric Random Problems

For the nonsymmetric problem two sets of dual variables  $u_i$ 's and  $v_i$ 's must be used. For this reason the minimal 1-tree calculation usually consumes more effort. Since this calculation is repeated continuously during the ascent scheme, we may expect that the overall computational effort will be larger than the symmetric case. Also the increased complexity of the dual space makes it more difficult for the ascent scheme to reach the true maximum of the dual function.

TABLE V.2

Problem Size	Dual objective before ascent	Dual objective after ascent	Time for ascent (seconds)	Primal Solution	%age theoretical deviation from lower bound	%age actual deviation from lower bound	Time for branch and bound (seconds)	Total computational time (seconds)
40	7956	10573	8.9	10573*	1.0%	0.0%	0.8	9.7
40	7998	10615	7.8	10615*	1.0%	0.0%	1.1	8.9
50	7721	9462	18.0	9533	1.0%	0.8%	7.5	25.5
50	7819	9495	16.0	9496	1.0%	0.0%	1.1	17.1
60	8765	10643	20.0	10643*	1.0%	0.0%	1.7	21.7
60	8779	10528	23.1	10528*	1.0%	0.0%	1.6	24.7
70	11089	13287	35.0	13368	2.0%	0.6%	12.4	47.4
70	11255	13418	28.2	13603	2.0%	1.2%	15.1	43.3
80	11764	14365	41.8	14671	5.0%	2.1%	4.1	45.9
80	11707	14199	55.2	14521	5.0%	2.2%	7.0	62.2

TABLE V.3

Problem Size	Dual objective before ascent	Dual objective after ascent	Time for ascent (seconds)	Primal Solution	%age actual deviation from lower bound	Time for branch and bound (seconds)	Total computational time (seconds)
40	7956	10573	8.9	10573*	0.0%	1.5	10.4
40	7998	10615	7.8	10615*	0.0%	0.9	8.7
50	7721	9462	18.0	9533	0.8%	15.6	33.6
50	7819	9495	16.0	9496	0.0%	1.1	17.1
60	8765	10643	20.0	10643*	0.0%	0.5	20.5
60	8779	10528	23.1	10528*	0.0%	1.5	24.6
70	11089	13287	35.0	13368	0.6%	17.0	52.0
70	11255	13418	28.2	13501	0.6%	77.9	106.1
80	11764	14365	41.8	14436	0.5%	4.0	45.8
80	11707	14199	55.2	14678**	3.2%	8.0	63.2

\*\*In this particular run the algorithm was not able to reduce the gap between the lower and upper bounds to 100 in 5 minutes, even though at least 96.8% quality solution is obtained after approximately 1 minute.

Computational results with random problems with  $n$  varying from 30 to 60 are demonstrated in the following two tables. For larger problems, both methods 1 and 2 were not able to reduce the gap between the lower and upper bounds to within 4 percent of the value of the upper bound in 5 minutes.

#### Method 1

Stepped fathoming 92%, 96%, and then 99% is employed for the first three problems whereas stepped fathoming 92% then 96% is used for the remainder of the problems. The results are shown in Table V.4.

#### Method 2

The same problems summarized above were run using method 2 of stepped fathoming where the interval of uncertainty is halved at each iteration. The process is terminated when the difference between the lower and upper bounds is less than  $\epsilon$ . The results are summarized in Table V.5.

From Tables V.4 and V.5 we see that methods 1 and 2 are comparable both from the quality of solution and computational time.

It is worthwhile noting that the procedure was not able to solve non-symmetric problems with size 70 and 80 cities within 5 minutes, whereas other procedures in the literature are able to do so (see for example Bellmore and Malone [3]). Also, good quality solutions for nonsymmetric problems seem harder to get than those for symmetric problems.

Table V.4

Problem Size	Dual objective before ascent	Dual objective after ascent	Time for ascent (seconds)	Primal Solution	%age theoretical deviation from lower bound	%age actual deviation from lower bound	Time for branch and bound (seconds)	Total computational time (seconds)
30	4899	7427	10.5	7519	1%	1 %	26.8	37.2
30	4899	7427	10.4	7519	1%	1 %	26.7	37.1
40	6212	9383	19.2	9606*	1%	2.3%	11.4	30.6
40	6212	9383	19.9	9724	4%	3.4%	1.2	21.1
50	7448	9881	28.1	10059	4%	1.8%	12.5	
50	7446	9879	24.2	10057	4%	1.8%	14.1	38.3
60	8134	10871	31.2	11309	4%	3.9%	3.9	35.1
60	8132	10869	29.8	11308	4%	3.9%	3.3	33.1

\*In this particular run the algorithm was not able to finish the 99% fathoming in 2 minutes.

TABLE V.5

Problem Size	Dual ob- jective before ascent	Dual ob- jective after ascent	Time for ascent (seconds)	Primal Solution	Termina- tion criterion $\epsilon$	%age actual deviation from lower bound	Time for branch and bound (seconds)	Total computa- tional time (seconds)
30	4899	7427	12.4	7519	100	1.2%	13.7	26.1
30	4899	7427	11.3	7519	100	1.2%	13.8	25.1
40	6212	9383	19.6	9520	300	1.4%	38.5	58.1
40	6212	9383	20.0	9520	300	1.4%	39.2	59.2
50	7448	9881	26.2	10059	300	1.8%	49.6	75.8
50	7446	9879	26.8	10057	300	1.8%	51.5	78.3
60	8134	10871	31.2	11309	440	3.9%	1.9	33.1
60	8132	10869	28.3	11308	440	3.9%	1.6	29.9

### CONCLUSION

We have described a duality based procedure for solving the traveling salesman problem. The computational experience with symmetric problems seems quite promising. However, for nonsymmetric problems there seems to be room for further improvement.

Subgradient optimization was used in this study as the main procedure for maximizing the dual. The performance of other procedures such as ascent, steepest ascent, decomposition, and search methods, remains to be tested.



## REFERENCES

1. Bazaraa, M. S. and J. J. Goode, "A Survey of Various Tactics for Generating Lagrangian Multipliers in the Context of Lagrangian Duality," Georgia Institute of Technology, Atlanta, Georgia, August 1974.
2. Bazaraa, M. S. and A. N. Elshafei, "The Concept of Stepped Fathoming in Tree Search Algorithms," Georgia Institute of Technology, Atlanta, Georgia, June 1974.
3. Bellmore, M. and J. C. Malone, "Pathology of Traveling-Salesman Subtour-Elimination Algorithms," Operations Research, Vol. 19, pp. 278-307, 1971.
4. Dantzig, G. B., D. R. Fulkerson, and S. M. Johnson, "Solution of a Large Scale Traveling Salesman Problem," Operations Research, Vol. 2, pp. 393-410, 1954.
5. Dijkstra, E. W., "A Note on Two Problems in Connection with Graphs," Numerische Mathematik, Vol. 1, pp. 269-271, 1959.
6. Held, M. and R. M. Karp, "A Dynamic Programming Approach to Sequencing Problems," SIAM Journal on Applied Mathematics, Vol. 10, pp. 196-210, 1962.
7. Held, M. and R. M. Karp, "The Traveling Salesman Problem and Minimal Spanning Trees," Operations Research, Vol. 18, pp. 1138-1162, 1970.
8. Held, M. and R. M. Karp, "The Traveling Salesman Problem and Minimum Spanning Trees: Part II," Mathematical Programming, Vol. 1, pp. 6-25, 1971.
9. Held, M., P. Wolfe, and H. D. Crowder, "Validation of Subgradient Optimization," Mathematical Programming, Vol. 6, pp. 62-88, 1974.
10. Karg, L. L. and G. L. Thompson, "A Heuristic Approach to Solving Traveling Salesman Problems," Management Science, Vol. 10, pp. 225-248, 1964.
11. Kruskal, J. B., "On the Shortest Spanning Subtree of the Graph and the Traveling Salesman Problem," Proceedings of American Mathematical Society, Vol. 1, pp. 48-50, 1956.

## PERSONNEL

This section presents the personnel involved in this research and the contributions they have made. It also gives a list of papers related to this research.

### Principal Investigator

Dr. Mokhtar S. Bazaraa

Associate Professor

School of Industrial and Systems Engineering

Georgia Institute of Technology

### Contribution

Organization and coordination of the project. Development of the algorithms in Sections III, IV, and V. Review of the programming development.

### Graduate Research Assistants

#### 1. Virginia Ruth Kluge

School of Industrial and Systems Engineering

Georgia Institute of Technology

#### Contribution

Computer program of the initial solutions of the quadratic assignment and traveling salesman problems.

#### 2. Peter H. Geddes

School of Industrial and Systems Engineering

Georgia Institute of Technology

#### Contribution

Applications of the quadratic set covering problem, quadratic assignment problem, and related problems.

3. Sinan S. Tumer

School of Industrial and Systems Engineering  
Georgia Institute of Technology

Contribution

Programming of the quadratic set covering algorithm. Generation and running of problems.

Acknowledgement

Dr. Jamie J. Goode of the School of Mathematics, Georgia Institute of Technology, contributed to the development of the traveling salesman algorithms of Section V.

Dr. A. N. Elshafei of the Institute of Planning in Cairo, Egypt, contributed to the development of the quadratic assignment algorithms of Section IV.

List of Papers Related to Project

1. M. S. Bazaraa and A. N. Elshafei, "The Concept of Stepped Fathoming in Tree Search Algorithms," submitted for publication.
2. M. S. Bazaraa and A. N. Elshafei, "Heuristic and Exact Procedures of the Quadratic Assignment Problem," submitted for publication.
3. M. S. Bazaraa and J. J. Goode, "The Traveling Salesman Problem: A Duality Approach," submitted for publication.
4. M. S. Bazaraa, "Computerized Layout Design: A Branch and Bound Approach," submitted for publication.

## APPENDIX

This appendix gives a listing of the programs developed for solving quadratic set covering problems, quadratic assignment problems, and traveling salesman problems. The programs are organized into 10 files. A brief description of each file and its elements is given below. This is followed by a complete listing of all the programs.

### 1. QSC

A set of programs for solving quadratic set covering problems in the context of layout design by enumeration. The file consists of the following elements.

- a. QSC. ASSGN
- b. QSC. BBOUND
- c. QSC. COSTUN
- d. QSC. INASS
- e. QSC. INPUT
- f. QSC. MAIN
- g. QSC. NEWBOU
- h. QSC. OUTPUT
- i. QSC. TIME

### 2. BAB

A set of programs for solving quadratic set covering problems by branch and bound. The file consists of the following elements.

- a. BAB. BBOUND
- b. BAB. COST

### 3. QAP5

A set of programs for solving quadratic assignment problems by branch and bound. The file consists of the following elements.

- a. QAP5. MAIN
- b. QAP5. UPDATE
- c. QAP5. LWBND
- d. QAP5. ARRNG
- e. QAP5. TIMEOUT
- f. QAP5. DTAIN

4. QAP4

A set of programs for solving quadratic assignment problems by branch and bound. The file consists of the following elements.

- a. QAP4. MAIN
- b. QAP4. ARRNG
- c. QAP4. UPDATE
- d. QAP4. LWBND
- e. QAP4. MAX
- f. QAP4. DTAIN
- g. QAP4. TIMEOUT

5. QAP7

A set of programs for solving quadratic assignment problems by branch and bound. The file consists of the following elements.

- a. QAP7. MAIN
- b. QAP7. DTAIN
- c. QAP7. GARY
- d. QAP7. HADY
- e. QAP7. LWBND2

6. QAIN

A set of programs for providing a heuristic solution to the quadratic assignment problem. The file consists of the following elements.

- a. QAIN. MAIN
- b. QAIN. DTAIN

## 7. NEW QAP

A set of programs for solving the dual of the quadratic assignment problem by subgradient optimization. The file consists of the following elements.

- a. NEWQAP. MAIN
- b. NEWQAP. UVW
- c. NEWQAP. DTAIN
- d. NEWQAP. KNAP
- e. NEWQAP. HADY
- f. NEWQAP. GARY
- g. NEWQAP. X

## 8. QAP-DATA

This is a file consisting of the data for the quadratic assignment problems attempted. The file consists of the following elements.

- a. QAP-DATA. PROBLEM 1
- b. QAP-DATA. PROBLEM 2
- c. QAP-DATA. PROBLEM 3
- d. QAP-DATA. PROBLEM 4
- e. QAP-DATA. PROBLEM 5
- f. QAP-DATA. PROBLEM 6
- g. QAP-DATA. PROBLEM 7
- h. QAP-DATA. PROBLEM 8

## 9. FATHOM

A set of programs for solving symmetric traveling salesman problems with subgradient optimization and then branch and bound. The file consists of the following elements.

- a. FATHOM. COPY
- b. FATHOM. INPUT
- c. FATHOM. INRAND
- d. FATHOM. IO
- e. FATHOM. MAIN

- f. FATHOM. MINPT
- g. FATHOM. MINPT 2
- h. FATHOM. ORDER/NEW
- i. FATHOM. ORDER/OLD
- j. FATHOM. OUTPUT/BATCH
- k. FATHOM. OUTPUT/DEMAND
- l. FATHOM. SHOW/BATCH
- m. FATHOM. SHOW/DEMAND
- n. FATHOM. START
- o. FATHOM. TIMEOUT
- p. FATHOM. TREE/NEW
- q. FATHOM. TREE/OLD
- r. FATHOM. VOGEL

#### 10. NONSYM

A set of programs for solving nonsymmetric traveling salesman problems by subgradient optimization and then branch and bound. The file consists of the following elements.

- a. NONSYM. CMINP3
- b. NONSYM. INPUT
- c. NONSYM. INRAND
- d. NONSYM. IO
- e. NONSYM. MAIN
- f. NONSYM. MINPT
- g. NONSYM. MINPT2
- h. NONSYM. ORDER/OLD
- i. NONSYM. OUTPUT/BATCH
- j. NONSYM. OUTPUT/DEMAND
- k. NONSYM. SHOW/BATCH
- l. NONSYM. SHOW/DEMAND
- m. NONSYM. START
- n. NONSYM. TIMEOUT
- o. NONSYM. TREE

[illegible]

FROM TIRON, 011A0930, BARDIN, 10, 1000  
CHGO CHG AND EXCESS LINES TO E-24-200  
CELL, 1DL MAIN

ELT0T7 RL1B70 02/21-16:58:42-(,0)

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000001 000 QEL(,SI TUMER-S*QSC,ASSON,,,10112010112
000002 000 SUBROUTINE ASSON(KOD,COST)
000003 000 C
000004 000 C ROUTINE TRIES TO ASSIGN UNASSIGNED OBJECTS
000005 000 C KOD :=1 IF COMPLETE ASSIGNMENT HAS BEEN MADE
000006 000 C =-1 OTHERWISE
000007 000 C
000008 000 C
000009 000 C
000010 000 INCLUDE BLANK
000011 000 INCLUDE INPUT
000012 000 C BEGIN TO ASSIGN ELEMENTS INTO THE LOCATIONS
000013 000 C IELT=ELEMENT NUMBER
000014 000 C
000015 000 IELT=IAS+1
000016 000 NA=NAREA(IELT)
000017 000 122 IF(IELT.GT.1) IL=ILAST(IELT-1)
000018 000 IF(IELT.EQ.1) IL=0
000019 000 C CHECK IF THE ROOM CAN BE ASSIGNABLE
000020 000 C
000021 000 130 ICOL=IL+INI(IELT)
000022 000 DO 132 IN=1,NA
000023 000 IA=A(IN,ICOL)
000024 000 IF(IA.EQ.0) WRITE(IWT,9999)IFLT,YAS,ICOL,INI(IELT),NA
000025 000 IF(BOARD(IA).NE.0) GO TO 135
000026 000 132 CONTINUE

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```

000026 000 GO TO 136
000027 000 135 INI(IELT)=INI(IELT)+1
000028 000 IF(INI(IELT).LE.NPLOC(IELT)) GO TO 130
000029 000 C CHANGE THE ALTERNATIVE
000030 000 C
000031 000 INI(IELT)=1
000032 000 128 IELT=IELT-1
000033 000 IF(IELT.EQ.0) GO TO 133
000034 000 COST=COST-INCR(T(IELT))
000035 000 C CLEAR THE BOARD
000036 000 C
000037 000 NA=NAREA(IELT)
000038 000 ICO=ELCOL(IELT)
000039 000 DO 129 I=1,NA
000040 000 IA=A(I,ICO)
000041 000 129 BOARD(IA)=0
000042 000 IF(INI(IELT).LT.NPLOC(IELT)) GO TO 131
000043 000 INI(IELT)=1
000044 000 GO TO 128
000045 000 131 INI(IELT)=INI(IELT)+1
000046 000 GO TO 122
000047 000 133 KOD=-1
000048 000 RETURN
000049 000 C CALCULATE INCREMENT OF COST
000050 000 C
000051 000 136 IF(IELT.EQ.1) GO TO 139
000052 000 IAS=IELT-1
000053 000 CALL COSTA(CR,NA,IELT,ICOL)
000054 000 C COMPARE THE INCREMENT OF COST
000055 000 C
000056 000 INCR(T(IELT))=CR
000057 000 COST=COST+INCR(T(IELT))
000058 000 139 ELCOL(IELT)=ICOL
000059 000 C MAKE THE NEW ASSIGNMENT
000060 000 C
000061 000 ICO=ELCOL(IELT)
000062 000 DO 138 J=1,NA
000063 000 IA=A(J,ICO)
000064 000 138 BOARD(IA)=IELT
000065 000 IELT=IELT+1
000066 000 NA=NAREA(IELT)
000067 000 IF(IELT.LE.NELT) GO TO 122
000068 000 KOD=1
000069 000 IAS=IELT
000070 000 RETURN
000071 000 END
000072 000 DELT,SI TUMER-S*QSC.BROUND,,,152335102712
000073 000 SUBROUTINE BBOUND
000074 000 C
000075 000 C ROUTINE CALCULATES THE NEW ROUND
000076 000 C AND CHECKS WHETHER FURTHER IMPROVEMENT IS POSSIBLE
000077 000 C
000078 000 C COST1 :COST BETWEEN ASSIGNED OBJECTS
000079 000 C COST2 :COST BETWEEN UNASSIGNED AND ASSIGNED OB
000080 000 C COST3 :COST BETWEEN UNASSIGNED OBJECTS
000081 000 C ISAVE :SAVE AREA FOR STARTING LOCATIONS
000082 000 C

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000083 000      INCLUDE BLANK
000084 000      INCLUDE INPUT
000085 000      COST1=CSTAR
000086 000      C      REMOVE THE LAST ASSIGNED ELEMENT FROM THE BOARD
000087 000      C
000088 000      IO=3
000089 000      IELT=NELT+1
000090 000      100      IELT=IELT-1
000091 000      ICO=ELCOL(IELT)
000092 000      C      WRITE(IWT,1003) IELT,ICO
000093 000      NA=NAAREA(IELT)
000094 000      NP=NPLOC(IELT)
000095 000      DO 105 I=1,NA
000096 000      IA=A(I,ICO)
000097 000      105      BOARD(IA)=0
000098 000      C      WRITE(IWT,1004)(A(I,ICO),I=1,NA)
000099 000      IF(IELT.EQ.1) IL=0
000100 000      IF(IELT.GT.1) IL=ILAST(IELT-1)
000101 000      IPOS=ICO-IL
000102 000      NOTAS(IELT,IPOS)=.TRUE.
000103 000      IAS=IAS-1
000104 000      C      COST1
000105 000      C
000106 000      COST1=COST1-INCR(IELT)
000107 000      C      CALCULATE A NEW BOUND AFTER DELETE
000108 000      C
000109 000      101      CALL NEWBOU(COST1,BOUND,KOD)
000110 000      IF(KOD.EQ.-1) GO TO 102
000111 000      IF(BOUND.LT.CSTAR) GO TO 104
000112 000      GO TO 102
000113 000      C      CHECK WHETHER IT IS PHYSICALLY POSSIBLE
000114 000      C      TO ASSIGN THE DELETED ELEMENT SOMEWHERE ELSE
000115 000      C
000116 000      104      DO 106 I=1,NP
000117 000      ICOL=IL+I
000118 000      DO 107 II=1,NA
000119 000      IA=A(II,ICOL)
000120 000      IF(BOARD(IA).NE.0.OR.NOTAS(IELT,II)) GO TO 106
000121 000      IPOS=I
000122 000      107      CONTINUE
000123 000      C      ASSIGN THE ELEMENT TO ITS NEW POSITION
000124 000      C
000125 000      CALL COSTA(CR,NA,IELT,ICOL)
000126 000      INCR(IELT)=CR
000127 000      COST1=COST1+CR
000128 000      ELCOL(IELT)=ICOL
000129 000      C      WRITE(IWT,1005) IELT,I,CR
000130 000      DO 108 IT=1,NA
000131 000      IA=A(IT,ICOL)
000132 000      108      BOARD(IA)=IELT
000133 000      C      WRITE(IWT,1004)(A(IT,ICOL),IT=1,NA)
000134 000      IAS=IAS+1
000135 000      C      END OF THE ELEMENTS
000136 000      C
000137 000      IF(IAS.LT.NELT) GO TO 110
000138 000      IF(CSTAR.GT.COST1) GO TO 109
000139 000      C      WRITE(IWT,1002)COST1,CSTAR

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000140      000      IELT=IELT+1
000141      000      GO TO 100
000142      000  109  CSTAR=COST1
000143      000      CALL OUTPUT(I0)
000144      000      I0=0
000145      000      IELT=IELT+1
000146      000      GO TO 100
000147      000  106  CONTINUE
000148      000  C      FREE THE LOCATIONS
000149      000  C
000150      000  102  IF(IELT.EQ.1.AND.IPOS.EQ.ILAST(1)) RETURN
000151      000      IF(IELT.GT.1) GO TO 111
000152      000      NOTAS(1,IPOS)=.TRUE.
000153      000      GO TO 104
000154      000  111  DO 103 I=1,NP
000155      000  103  NOTAS(IELT,I)=.FALSE.
000156      000      GO TO 100
000157      000  C
000158      000  C      CHECK FOR A BETTER ROOM
000159      000  C
000160      000  110  CALL NEWROU(COST1,BOUND,KOD)
000161      000      IF(KOD.EQ.-1) GO TO 112
000162      000      IF(BOUND.GE.CSTAR) GO TO 112
000163      000  C      TRY TO ASSIGN THE NEXT ELEMENT
000164      000  C
000165      000      IELT=IELT+1
000166      000      NP=NPLC(IELT)
000167      000      NA=NAREA(IELT)
000168      000      IL=ILAST(IELT-1)
000169      000      GO TO 104
000170      000  C112 WRITE(IWT,1006) IELT,IPOS
000171      000  112  NOTAS(IELT,IPOS)=.TRUE.
000172      000      DO 113 IT=1,NA
000173      000      IA=A(IT,ICOL)
000174      000  113  BOARD(IA)=0
000175      000  C      WRITE(IWT,1004) (A(IT,ICOL),IT=1,NA)
000176      000      COST1=COST1-INCR(IELT)
000177      000      IAS=IAS-1
000178      000      GO TO 104
000179      000  1002 FORMAT(/,2X,'NEW COST=',F12.2,' OPTIMUM COST=',F12.2)
000180      000  1003 FORMAT(/,2X,'ROOM ',I3,' IS DELETED FROM POSITION:',I4,
000181      000      */2X,'BLOCKS:')
000182      000  1004 FORMAT(2X,20I5)
000183      000  1005 FORMAT(/,2X,'ROOM ',I4,' IS ASSIGNED TO POSITION',I4,
000184      000      */2X,' WITH COST:',F10.2/2X,'BLOCK')
000185      000  1006 FORMAT(/,2X,'ASSIGN ROOM ',I4,' PRESENTLY IN POSITION',I4,
000186      000      */2X,' TO NEXT FREE POSITION',/2X,'BLOCKS',/)
000187      000      END
000188      000  QELT,SI TUMER-S*QSC.COSTA,,,141302101112
000189      000      SUBROUTINE COSTA(CR,NA,IELT,ICOL)
000190      000  C      INCREMENTAL COST BETWEEN ASSIGNED ELEMENTS
000191      000      INCLUDE BLANK
000192      000      CR=0.
000193      000      R2=NA
000194      000      DO 133 IIE=1,IAS
000195      000      R1=NAREA(IIE)
000196      000      R3=IWEIG(IELT,IIE)

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000197      000  C      FIND UNIT COMMUNICATION AND INCREMENT OF COST
000198      000  C
000199      000      ICO=ELCOL(IIE)
000200      000      UNIT=R3/(R1*R2)
000201      000      INA=NAREA(IIE)
000202      000      DO 134 JE=1,INA
000203      000      I1=A(JE,ICO)
000204      000      DO 134 JM=1,NA
000205      000      I2=A(JM,ICOL)
000206      000      CR=CR+IDIST(I1,I2)*UNIT
000207      000      134 CONTINUE
000208      000      RETURN
000209      000      END
000210      000  BELT,SI TUMER-S*QSC.COSTUM,,,150750101112
000211      000      SUBROUTINE COSTUM(ICON,INA,IUNAS,ICOL,KOD)
000212      000  C
000213      000  C      ROUTINE CALCULATES THE COST
000214      000  C      BETWEEN UNASSIGNED ELEMENTS
000215      000  C
000216      000      INCLUDE BLANK
000217      000      INCLUDE INPOUT
000218      000      BIG=1.0E38
000219      000      IBEG=IUNAS+1
000220      000      R2=INA
000221      000      COST=0.
000222      000      DO 130 IEL=IBEG,NELT
000223      000      KOD=-1
000224      000      SAVE=BIG
000225      000      INA=NAREA(IEL)
000226      000      NP=NPLOC(IEL)
000227      000      IL=ILAST(IEL-1)
000228      000      R1=INA
000229      000      R3=IWEIG(IUNAS,IEL)
000230      000      UNIT=R3/(R1*R2)
000231      000      DO 125 ILOC=1,NP
000232      000      ICO=ILOC+IL
000233      000  C      CHECK WHETHER ASSIGNMENT CAN BE MADE
000234      000  C
000235      000      DO 120 IAREA=1,INA
000236      000      IA=A(IAREA,ICO)
000237      000      DO 121 I=1,NA
000238      000      IF(IA.EQ.A(I,ICOL)) GO TO 125
000239      000      121 CONTINUE
000240      000      IF(IA.EQ.0) WRITE(JWT,999)IUNAS,IEL,ICOL,ICO,NA,INA
000241      000      IF(BOARD(IA).NE.0.OR.NOTAS(IFL,ILOC)) GO TO 125
000242      000      120 CONTINUE
000243      000      KOD=1
000244      000      CR=0.
000245      000  C      CALCULATE COST BETWEEN OBJECTS.
000246      000  C
000247      000      DO 134 JE=1,INA
000248      000      DO 134 JM=1,NA
000249      000      I1=A(JE,ICO)
000250      000      I2=A(JM,ICOL)
000251      000      134 CR=CR+IDIST(I1,I2)*UNIT
000252      000      IF(SAVE.GT.CR) SAVE=CR
000253      000      125 CONTINUE

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000254 000 IF(KOD.EQ.-1) RETURN
000255 000 COST=COST+SAVE
000256 000 130 CONTINUE
000257 000 RETURN
000258 000 END
000259 000 DELT,SI TUMER-S*QSC.DISTCE,,,113117062412
000260 000 SUBROUTINE DISTCE(KOD)
000261 000 INCLUDE BLANK
000262 000 INCLUDE INPUT
000263 000 C ROUTINE WHICH CALCULATES IDISTANCES ON THE BOARD
000264 000 INTEGER X(1BLOCK),Y(1BLOCK)
000265 000 N=NBLK
000266 000 READ(IRD,1000) (X(I),Y(I),I=1,N)
000267 000 1000 FORMAT(16I5)
000268 000 N1=N-1
000269 000 GO TO (100,200,300),KOD
000270 000 C RECTILINEAR IDISTANCES KOD=1
000271 000 C
000272 000 100 DO 110 I=1,N1
000273 000 L=I+1
000274 000 DO 110 J=L,N
000275 000 IDIST(I,J)=IABS(X(I)-X(J))+IABS(Y(I)-Y(J))
000276 000 110 IDIST(J,I)=IDIST(I,J)
000277 000 GO TO 999
000278 000 C EUCLIDEAN IDISTANCES KOD=2
000279 000 C
000280 000 200 DO 210 I=1,N1
000281 000 L=I+1
000282 000 DO 210 J=L,N
000283 000 IDIST(I,J)=SQRT((X(I)-X(J))**2+(Y(I)-Y(J))**2)
000284 000 210 IDIST(J,I)=IDIST(I,J)
000285 000 GO TO 999
000286 000 C EUCLIDEAN SQUARE IDISTANCE
000287 000 C
000288 000 300 DO 310 I=1,N1
000289 000 L=I+1
000290 000 DO 310 J=L,N
000291 000 IDIST(I,J)=(X(I)-X(J))**2+(Y(I)-Y(J))**2
000292 000 310 IDIST(J,I)=IDIST(I,J)
000293 000 C OUTPUT DISTANCE MATRIX
000294 000 C
000295 000 C999 WRITE(IWT,1001)
000296 000 C DO 998 I=1,N
000297 000 C998 WRITE(IWT,1000) (IDIST(I,J),J=1,N)
000298 000 C1001 FORMAT(/6X,'DISTANCE MATRIX',/)
000299 000 999 RETURN
000300 000 END
000301 000 DELT,SI TUMER-S*QSC.FORS,,,103724102312
000302 000 QFOR,S QSC.ASSGN
000303 000 QFOR,S QSC.COSTA
000304 000 QFOR,S QSC.DISTCE
000305 000 QFOR,S QSC.INASS
000306 000 QFOR,S QSC.INPUT
000307 000 QFOR,S QSC.NEWBOU
000308 000 QFOR,S QSC.MAIN
000309 000 QFOR,S QSC.OUTPUT
000310 000 QFOR,S QSC.TIME

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000311 000 QFOR,S QSC.BBOUND
000312 000 QFOR,S QSC.COSTUN
000313 000 QEOF
000314 000 DELT,SI TUMER-S*QSC.FORY,,,137524101112
000315 000 QFOR,Y QSC.ASSGN
000316 000 QFOR,Y QSC.COSTA
000317 000 QFOR,Y QSC.DISTCE
000318 000 QFOR,Y QSC.INASS
000319 000 QFOR,Y QSC.INPUT
000320 000 QFOR,Y QSC.NEWROU
000321 000 QFOR,Y QSC.MAIN
000322 000 QFOR,Y QSC.OUTPUT
000323 000 QFOR,Y QSC.TIME
000324 000 QFOR,Y QSC.BBOUND
000325 000 QFOR,Y QSC.COSTUN
000326 000 QEOF
000327 000 DELT,SI TUMER-S*QSC.INASS,,,137524102612
000328 000 SUBROUTINE INASS(IHAR)
000329 000 C MAKE THE INITIAL ASSIGNMENT ON THE BOARD
000330 000 INCLUDE BLANK
000331 000 INCLUDE INPUT
000332 000 IO=1
000333 000 ILAST(1)=NPLOC(1)
000334 000 DO 117 I=2,NELT
000335 000 117 ILAST(I)=ILAST(I-1)+NPLOC(I)
000336 000 CSTAR=1.0E38
000337 000 DO 119 I=1,NELT
000338 000 119 INI(I)=1
000339 000 IAS=0
000340 000 C ASSIGN THE OBJECTS
000341 000 C
000342 000 DO 122 I=1,NBLK
000343 000 122 BOARD(I)=0
000344 000 COST=0.
000345 000 121 CALL ASSGN(KOD,COST)
000346 000 C COMPARE THE PREVIOUS COST FOR IMPROVEMENT
000347 000 C
000348 000 IF(KOD.EQ.-1) GO TO 994
000349 000 MI=MI+1
000350 000 IF(CSTAR.LE.COST) GO TO 140
000351 000 CSTAR=COST
000352 000 C OUTPUT THE ASSIGNMENTS
000353 000 C
000354 000 145 CALL OUTPUT(IO)
000355 000 IF(IO.EQ.1) IO=2
000356 000 C CHECK TIME FOR FURTHER IMPROVEMENT
000357 000 C
000358 000 140 IT1=ITIME(IT2,IT3)
000359 000 T=IT2/5000.
000360 000 IF(T.GT.MAXTIM) RETURN
000361 000 C SEARCH FOR BETTER INITIAL SOLUTION
000362 000 C
000363 000 155 INI(IAS)=INI(IAS)+1
000364 000 COST=COST-INCR(IAS)
000365 000 NA=NAREA(IAS)
000366 000 ICO=ELCOL(IAS)
000367 000 DO 150 I=1,NA

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000368 000 IA=A(I,ICO)
000369 000 150 BOARD(IA)=0
000370 000 IAS=IAS-1
000371 000 IF(IAS.LT.0) GO TO 994
000372 000 IF(INI(IAS+1).LE.NPLOC(IAS+1)) GO TO 121
000373 000 INI(IAS+1)=1
000374 000 GO TO 155
000375 000 994 MI=MI+1
000376 000 IF(MI.EQ.1) GO TO 995
000377 000 WRITE(IWT,101)
000378 000 ISAB=1
000379 000 RETURN
000380 000 995 WRITE(IWT,100)
000381 000 STOP
000382 000 100 FORMAT(/// STOP** THERE IS NO POSSIBILITY TO MAKE COMPLETE *
000383 000 *,'INITIAL ASSIGNMENT')
000384 000 101 FORMAT(/// IT IS NOT POSSIBLE TO MAKE FURTHER *
000385 000 *,'IMPROVEMENT')
000386 000 END
000387 000 DELT,SI TUMER-S*QSC.INPUT,,,157086071012
000388 000 SUBROUTINE INPUT
000389 000 C IDIS :=1 DISTANCE WILL BE CALCULATED
000390 000 C =0 DISTANCE WILL BE READ
000391 000 INCLUDE BLANK
000392 000 INCLUDE INPUT
000393 000 C READ DATA
000394 000 C
000395 000 READ(IRD,1001) NELT,NBLK,MAXIM,IDIS
000396 000 C ELEMENT CARDS
000397 000 C
000398 000 DO 104 I=1,NELT
000399 000 READ(IRD,1002) M,NPLOC(M),NAREA(M),NLIM(M)
000400 000 IF(I.NE.M) GO TO 997
000401 000 104 CONTINUE
000402 000 C LIMITATION CARDS
000403 000 C
000404 000 DO 102 IE=1,NELT
000405 000 NL=NLIM(IE)
000406 000 IF(NL.EQ.0) GO TO 102
000407 000 READ(IRD,1001)M,(LIMIT(M,I),I=1,NL)
000408 000 IF(IE.NE.M.OR.LIMIT(M,NL).EQ.0) GO TO 998
000409 000 102 CONTINUE
000410 000 C CREATE A MATRIX
000411 000 IC=0
000412 000 DO 110 I=1,NELT
000413 000 NP=NPLOC(I)
000414 000 NA=NAREA(I)
000415 000 DO 110 J=1,NP
000416 000 IC=IC+1
000417 000 READ(IRD,1001) M,(A(II,IC),II=1,NA)
000418 000 IF(I.NE.M.OR.A(NA,IC).EQ.0.OR.A(NA+1,IC).NE.0) GO TO 999
000419 000 110 CONTINUE
000420 000 C READ DISTANCES BETWEEN EACH BLOCK
000421 000 C
000422 000 IF(IDIS.EQ.0) GO TO 113
000423 000 CALL DISTCE(1)
000424 000 GO TO 114

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000425 000
000426 000 113 NBL=NBLK-1
000427 000 DO 111 I=1,NBL
000428 000 K=I+1
000429 000 READ(IRD,1001) M,(IDIST(M,MJ),MJ=K,NBLK)
000430 000 IF(I.NE.M) GO TO 995
000431 000 DO 111 J=1,NBLK
000432 000 IDIST(J,I)=IDIST(I,J)
000433 000 C READ WEIGHT BETWEEN ELEMENTS
000434 000 C
000435 000 114 NLT=NELT-1
000436 000 DO 112 I=1,NLT
000437 000 K=I+1
000438 000 READ(IRD,1001) M,(IWEIG(M,IJ),IJ=K,NELT)
000439 000 IF(I.NE.M) GO TO 996
000440 000 DO 112 J=1,NELT
000441 000 IWEIG(J,I)=IWEIG(I,J)
000442 000 112 RETURN
000443 000 C ERROR MESSAGES
000444 000 C
000445 000 995 WRITE(IWT,1006) I,M
000446 000 STOP
000447 000 996 WRITE(IWT,1007) I,M
000448 000 STOP
000449 000 997 WRITE(IWT,1003) I,M
000450 000 STOP
000451 000 998 WRITE(IWT,1004) IE,M
000452 000 STOP
000453 000 999 WRITE(IWT,1005) I,M
000454 000 STOP
000455 000 C FORMAT STATEMENTS
000456 000 1001 FORMAT(13I5)
000457 000 1002 FORMAT(4I5)
000458 000 1003 FORMAT(/// STOP** CHECK ELEMENT CARD:',I5,2X,(M=',I5)
000459 000 1004 FORMAT(/// STOP** CHECK LIMITATION CARD:',I5,2X,'M=',I5)
000460 000 1005 FORMAT(/// STOP** CHECK ELEMENT:',I5,' OF MATRIX A M=',I5)
000461 000 1006 FORMAT(/// STOP** CHECK DISTANCE MATRIX ELEMENT:',I5,' M=',I5)
000462 000 1007 FORMAT(/// STOP** CHECK WEIGHT MATRIX ELEMENT:',I5,' M=',I5)
000463 000 END
000464 000 DELT,SI TUMER-S*QSC.MAIN,,,140721101112
000465 000 C
000466 000 C
000467 000 C QUADRATIC SET COVERING PROBLEM
000468 000 C
000469 000 C
000470 000 C VARIABLES:
000471 000 C
000472 000 C A :BINARY MATRIX WHICH SHOWS POSSIBLE LOCATIONS OF ELEMENTS.
000473 000 C ASRANK:ASSIGNMENT ORDER
000474 000 C AUXBOA:AUXILIARY ASSIGNMENT BOARD FOR MODIFICATION
000475 000 C BOARD :ASSIGNMENT BOARD
000476 000 C CSTAR :OPTIMAL COST (C1+C2+C3)
000477 000 C ELCOL :THE BEST CHOICE OF LOCATION OF ELEMENT I
000478 000 C IAS :NUMBER OF ALREADY ASSIGNED ELEMENTS
000479 000 C ICOL :COLUMN NUMBER OF MATRIX A
000480 000 C IDIST :DISTANCE FROM BLOCK I TO BLOCK J
000481 000 C IELT :ELEMENT NUMBER

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000482 000 C      ILAST :LAST COLUMN OF POSSIBLE LOCATIONS OF EACH ELEMENT IN A MATRIX
000483 000 C      INI   :INITIAL BEGINNING COLUMN FOR TRIALS
000484 000 C      INCR1 :INCREMENT OF COST AFTER ASSIGNMENT
000485 000 C      IUNIT :UNIT COMMUNICATION BETWEEN THE BLOCKS OF THE ELEMENTS
000486 000 C      IWEIG :WEIGHT BETWEEN ELEMENTS
000487 000 C      IFIRST:ELEMENT WHICH THE FIRST ASSIGNMENT WAS DONE
000488 000 C      LIMIT :ADJACENT ASSIGNMENT LIMITATION
000489 000 C      MAXTIM:MAXIMUM TIME IN SECONDS TILL THE INITIAL SOLUTION
000490 000 C              WILL BE IMPROVED
000491 000 C      NAREA :NUMBER OF BLOCKS WHICH AN ELEMENT OCCUPIES
000492 000 C      NBLK  :NUMBER OF BLOCKS
000493 000 C      NELT  :NUMBER OF ELEMENTS
000494 000 C      NLIM  :NUMBER OF LIMITATIONS ON EACH ELEMENT
000495 000 C      NPLOC :NUMBER OF POSSIBLE LOCATIONS WHERE AN ELEMENT CAN BE LOCATED
000496 000 C
000497 000      INCLUDE BLANK
000498 000      INCLUDE INPOT
000499 000 C
000500 000 C      READ DATA
000501 000      CALL INPUT
000502 000 C      MAKE THE INITIAL ASSIGNMENT
000503 000      CALL INASS(IBAB)
000504 000      IF (IBAB.EQ.1) STOP
000505 000      CALL BBOUND
000506 000      END
000507 000 @FLT,SI TUMER-S*QSC.MAPN-XQT,,,140054101112
000508 000 @MAP,N
000509 000 IN QSC.MAIN
000510 000 IN QSC.INPUT
000511 000 IN QSC.INASS
000512 000 IN QSC.COSTA
000513 000 IN QSC.NEWBOU
000514 000 IN QSC.ASSGN
000515 000 IN QSC.DISTCE
000516 000 IN QSC.OUTPUT
000517 000 IN QSC.TIME
000518 000 IN QSC.DBOUND
000519 000 IN QSC.COSTUN
000520 000 LIB SYSTEMS*MATHSTAT.
000521 000 END
000522 000 @XQT
000523 000 @FLT,SI TUMER-S*QSC.MAPS-XQT,,,104266102312
000524 000 @MAP,S
000525 000 IN QSC.MAIN
000526 000 IN QSC.INPUT
000527 000 IN QSC.INASS
000528 000 IN QSC.COSTA
000529 000 IN QSC.NEWBOU
000530 000 IN QSC.ASSGN
000531 000 IN QSC.DISTCE
000532 000 IN QSC.OUTPUT
000533 000 IN QSC.TIME
000534 000 IN QSC.DBOUND
000535 000 IN QSC.COSTUN
000536 000 LIB SYSTEMS*MATHSTAT.
000537 000 END
000538 000 @XQT

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000539 000 BELT,SI TUMER-S*QSC.NEWBOU,,,100705102712
000540 000 SUBROUTINE NEWBOU(COST1,BOUND,KOD)
000541 000 C++ROUTINE CALCULATES THE NEW BOUND
000542 000 C
000543 000 INCLUDE BLANK
000544 000 INCLUDE INPOUT
000545 000 BIG=1.E38
000546 000 COST23=0.
000547 000 IB=IAS+1
000548 000 DO 130 IUNAS=IB,NELT
000549 000 KOD1=-1
000550 000 KOD2=-1
000551 000 SAVE=BIG
000552 000 NA=NAREA(IUNAS)
000553 000 NP=NPLOC(IUNAS)
000554 000 IF(IUNAS.EQ.1) IL=0
000555 000 IF(IUNAS.GT.1) IL=ILANT(IUNAS-1)
000556 000 DO 125 ILOC=1,NP
000557 000 ICOL=ILOC+IL
000558 000 C CALCULATE COST2 & COST3
000559 000 C
000560 000 DO 120 IAREA=1,NA
000561 000 IA=A(IAREA,ICOL)
000562 000 IF(IA.EQ.0) WRITE(IWT,9999) IUNAS,IAS
000563 000 C CHECK WHETHER ASSIGNMENT CAN BE MADE
000564 000 C
000565 000 IF(BOARD(IA).NE.0.OR.NOTAS(IUNAS,ILOC)GO TO 125
000566 000 120 CONTINUE
000567 000 KOD1=1
000568 000 CR=0.
000569 000 CU=0.
000570 000 C COST2
000571 000 C
000572 000 IF(IAS.EQ.0) GO TO 121
000573 000 CALL COSTA(CR,NA,IUNAS,ICOL)
000574 000 C COST3
000575 000 C
000576 000 121 IF(IUNAS.EQ.NELT) GO TO 124
000577 000 CALL COSTUN(CU,NA,IUNAS,ICOL,KOD)
000578 000 IF(KOD.EQ.-1) GO TO 125
000579 000 124 C23=CR+CU
000580 000 IF(SAVE.GT.C23) SAVE=C23
000581 000 KOD2=1
000582 000 125 CONTINUE
000583 000 IF(KOD1.EQ.-1.OR.KOD2.EQ.-1) GO TO 102
000584 000 C
000585 000 COST23=COST23+SAVE
000586 000 130 CONTINUE
000587 000 BOUND=COST1+COST23
000588 000 C WRITE(IWT,1000) IAS
000589 000 C WRITE(IWT,1001) BOUND,COST1,COST2,COST3
000590 000 C CALL TIME
000591 000 RETURN
000592 000 C102 WRITE(IWT,1002) IAS
000593 000 102 KOD=-1
000594 000 RETURN
000595 000 1000 FORMAT(/,2X,'FOR ',I2,' ASSIGNED ELEMENTS BRANCH AND',

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000596 000      *' BOUND RESULTS:')
000597 000 1001 FORMAT(/2X,'NEWBOUND=',F11.2,' COST1=',F11.2,' COST2=',F11.2,
000598 000      *,' COST3=',F11.2)
000599 000 1002 FORMAT(/2X,'FOR',I4,' ASSIGNED ELEMENTS BRANCH AND BOUND',
000600 000      *' PHYSICALLY IS NOT POSSIBLE')
000601 000      END
000602 000 QFLT,SI TUMER-S*QSC.OUTPUT,,,152177102712
000603 000      SUBROUTINE OUTPUT(M)
000604 000 C          OUTPUT THE RESULTS
000605 000      INCLUDE BLANK
000606 000      INCLUDE INPOUT
000607 000      IF(M.EQ.0) GO TO 110
000608 000      IF(M.EQ.1) WRITE(IWT,1010)
000609 000      IF(M.EQ.2) WRITE(IWT,1012) MAXTIM
000610 000      IF(M.EQ.3) WRITE(IWT,1013)
000611 000      IF(M.EQ.2) M=0
000612 000 110      WRITE(IWT,1009) CSTAR
000613 000      WRITE(IWT,1011)
000614 000      DO 149 I=1,NELT
000615 000      IA=ELCOL(I)
000616 000      NA=NAREA(I)
000617 000 149      WRITE(IWT,1001) I,(AIJ,IA),J=1,NA)
000618 000      CALL TIME
000619 000      RETURN
000620 000 1001 FORMAT(16I5)
000621 000 1009 FORMAT(/2X,'*** COST=',F10.2)
000622 000 1010 FORMAT(///2X,'INITIAL ASSIGNMENTS'/2X,19(' '))
000623 000 1011 FORMAT(/2X,'ROOM',2X,'LOCATIONS ON THE BOARD -----')
000624 000 1012 FORMAT(///2X,'IMPROVED INITIAL ASSIGNMENTS'/2X,27(' ')/
000625 000      *2X,'IMPROVEMENT TILL',I5,' SECONDS')
000626 000 1013 FORMAT(/2X,'BRANCH AND BOUND IMPROVEMENT'/2X,29(' '))
000627 000      END
000628 000 QFLT,SI TUMER-S*QSC.RFOR,,,137545101112
000629 000 QRFUR,CS QSC.ASSGN
000630 000 QRFUR,CS QSC.COSTA
000631 000 QRFUR,CS QSC.DISTCE
000632 000 QRFUR,CS QSC.INASS
000633 000 QRFUR,CS QSC.INPUT
000634 000 QRFUR,CS QSC.NEWBOU
000635 000 QRFUR,CS QSC.MAIN
000636 000 QRFUR,CS QSC.OUTPUT
000637 000 QRFUR,CS QSC.TIME
000638 000 QRFUR,CS QSC.BROUND
000639 000 QRFUR,CS QSC.COSTUN
000640 000 QEOF
000641 000 QFLT,SI TUMER-S*QSC.TIME,,,157000061312
000642 000      SUBROUTINE TIME
000643 000      INCLUDE INPOUT
000644 000      K=ITIME(I,J)
000645 000      T=I/5000.
000646 000      WRITE(IWT,2000)T
000647 000 2000 FORMAT(' TIME',F10.3,' SEC')
000648 000      RETURN
000649 000      END
000650 000 QPDP,FI TUMER-S*QSC.BLANK,,,154360101112
000651 000 BLANK      PROC
000652 000      PARAMETER IROOM=25,IBLOCK=80

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000653 000    PARAMETER ICLMN=200,IROWS=30
000654 000    COMMON NPLOC(IROOM),NAREA(IROOM),LIMIT(IROOM,IROOM),
000655 000    *NLIN(IROOM),IWEIG(IROOM,IROOM),DIST(IBLOCK,IBLOCK),
000656 000    *ILAST(IROOM),A(IROWS,ICLMT),ASRANK(IROOM),NOTAS(IROOM,
000657 000    *IROOM),INCR(IROOM),INIT(IROOM),BOARD(IBLOCK)
000658 000    *,ELCOL(IROOM),IAS,MAXTIM,ELT,FINR,
000659 000    *CSTAR
000660 000    INTEGER A,ASRANK,ELCOL,BOARD
000661 000    REAL INCR
000662 000    LOGICAL NOTAS
000663 000    END
000664 000    QPDP,FI TUMER-S*QSC.INPUT,.,,154402101112
000665 000    INPUT PROC
000666 000    PARAMETER IWT=6,IRD=5
000667 000    9999 FORMAT( )
000668 000    END
000669 000    QPDP,FI TUMER-S*QSC.BLANK,.,,154402101112
000670 000    BLANK PROC
000671 000    PARAMETER IROOM=25,IBLOCK=30
000672 000    PARAMETER ICLMN=200,IROWS=30
000673 000    COMMON NPLOC(IROOM),NAREA(IROOM),LIMIT(IROOM,IROOM),
000674 000    *NLIN(IROOM),IWEIG(IROOM,IROOM),DIST(IBLOCK,IBLOCK),
000675 000    *ILAST(IROOM),A(IROWS,ICLMT),ASRANK(IROOM),NOTAS(IROOM,
000676 000    *IROOM),INCR(IROOM),INIT(IROOM),BOARD(IBLOCK)
000677 000    *,ELCOL(IROOM),IAS,MAXTIM,ELT,FINR,
000678 000    *CSTAR
000679 000    INTEGER A,ASRANK,ELCOL,BOARD
000680 000    REAL INCR
000681 000    LOGICAL NOTAS
000682 000    END
000683 000    QPDP,FI TUMER-S*QSC.INPUT,.,,154402101112
000684 000    INPUT PROC
000685 000    PARAMETER IWT=6,IRD=5
000686 000    9999 FORMAT( )
000687 000    END
000688 000
000689 000    RUNID IE0023 FILE PUB000IE0023 PART NO 000 DATE 051774 UNIT CP1 IE0023
000690 000    RELT,SI TUMER-S*SINAN-BAB.BBOUND,.,,170256041312
000691 000    C
000692 000    C
000693 000    C
000694 000    C    B R A N C H   A N D   B O U N D
000695 000    C
000696 000    C
000697 000    C    V A R I A B L E S :
000698 000    C    ASELT :LOCATION OF THE ASSIGNED ELEMENT
000699 000    C    ASRANK :ASSIGNMENT ORDER OF THE ELEMENTS
000700 000    C    ASSEL :ASSIGNED ELEMENTS
000701 000    C    AUX-I :AREAS USED FOR TEMPORARY CALCULATIONS
000702 000    C    AUXBOA:AUXILIARY ASSIGNMENT BOARD FOR MODIFICATION
000703 000    C    BOARD :ASSIGNMENT BOARD
000704 000    C    DISTFR:LOCATION WHICH THE DISTANCE MEASURED FROM
000705 000    C    DISTTO:LOCATION WHICH THE DISTANCE MEASURED TO
000706 000    C    ELEFR :ELEMENT WHICH THE CONNECTION LEAVES (FROM)
000707 000    C    ELETO :ELEMENT TO WHICH THE CONNECTION REACHES
000708 000    C    ELTCON:NO. OF CONNECTIONS FROM ELEMENT I TO ELEMENT J
000709 000    C    FINRAN:ELEMENTS IN DESCENDING ORDER ACCORDING TO

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000710	000	C	IAS :ASSIGNED ELEMENT NUMBER
000711	000	C	ICOL :ELEMENT NUMBER TO WHICH CONNECTION ENTERS
000712	000	C	IDIST :DISTANCE FROM LOCATION I TO LOCATION J
000713	000	C	IDISTS:LOCATIONS IN ASCENDING ORDER ACCORDING
000714	000	C	TO TOTAL DISTANCES FROM EACH LOCATION
000715	000	C	IELTC3:ARRAY WHICH THE ELEMENT CONNECTIONS ARE SORTED
000716	000	C	ILOCC3:ARRAY WHICH ELEMENT DISTANCES ARE SORTED
000717	000	C	IPERC :PERCENT OF ACCURACY DESIRED ON OPTIMUM SOLUTION
000718	000	C	ILO :LOCATION NUMBER WHICH THE ELEMENT HAS BEEN ASSIGNED
000719	000	C	IRE :ELEMENT NUMBER WHICH IS REMOVED
000720	000	C	IROW :ELEMENT NUMBER FROM WHICH CONNECTION LEAVES
000721	000	C	ITREE :BOARD WHICH SHOWS THE REMOVED ELEMENTS FROM TREE
000722	000	C	KCON :TOTAL NUMBER OF ELEMENT CONNECTIONS
000723	000	C	KDIS :TOTAL NUMBER OF DISTANCES BETWEEN LOCATIONS
000724	000	C	KODIS :=0 IF RECTILINEAR DISTANCES ARE DESIRED
000725	000	C	=1 IF EUCLIDEAN DISTANCES ARE DESIRED
000726	000	C	LOCOD :=0 IF DISTANCES ARE TO BE CALCULATED BY THE
000727	000	C	PROGRAM , ON THE ASSIGNMENT SHEET
000728	000	C	=1 IF DISTANCES ARE TO BE READ EXTERNALLY
000729	000	C	NCONS :ELEMENTS IN ASCENDING ORDER ACCORDING
000730	000	C	TO NUMBER OF CONNECTIONS TO OTHER ELEMENTS
000731	000	C	MELT :ASSIGNMENT ORDER
000732	000	C	NASSEL:NUMBER OF ASSIGNED ELEMENTS
000733	000	C	NDIM :MINIMUM DIMENSIONS OF ASSIGNMENT BOARD
000734	000	C	NELT :NUMBER OF ELEMENTS
000735	000	C	NLOC :NUMBER OF LOCATIONS
000736	000	C	NWIRES:ELEMENTS IN ASCENDING ORDER ACCORDING
000737	000	C	TO TOTAL NUMBER OF WIRES
000738	000	C	
000739	000	C	
000740	000		COMMON/AUXILI/AUX5(800),AUX6(800)
000741	000		COMMON/AREA/AUXBOA(40),AUX1(40),AUX2(40),AUX3(40),ITREE(35,40),
000742	000		1IDIST(40,40),ELTCUN(35,40),ASELT(35),ASSEL(35),MELT,NLOC,NASSEL
000743	000		2,KDIS,KCON,IELTC3(650),ILOCC3(600),ELEFR(650),DISTO(650),
000744	000		3DISTFR(800),DISTTO(800)
000745	000		DIMENSION ICOL(5),IVAL(5),IDISTS(40),NWIRES(35),NCONS(35)
000746	000		INTEGER ELTCUN,BOARD(40),EMPTY,FINISH(35),AUX1,AUX2,AUX3,
000747	000		1ASRANK(35),ASELT,AUXBOA,FROM,TO,CSTAR,CNEW,C1,C2,C3,ASSEL
000748	000		2,ELETO,ELEFR,DISTFR,DISTTO,AUX4(400),AUX5,AUX6,OPER
000749	000		IFIN=1
000750	000		IRD=5
000751	000		IWT=6
000752	000		INFIN=10E+10
000753	000		EMPTY=0
000754	000		READ(IRD,1001) NELT,IPERC,ICP,LOCOD,KODIS
000755	000		IF(LOCOD.EQ.0) GO TO 100
000756	000	C	READ DISTANCES FROM EACH LOCATION
000757	000	C	
000758	000		READ(IRD,1001) NLOC
000759	000		NLC=NLOC-1
000760	000		DO 96 I=1,NLC
000761	000		K=I+1
000762	000		READ(IRD,1003) (IDIST(I,J),J=K,NLOC)
000763	000		DO 96 J=1,NLOC
000764	000	96	IDIST(J,I)=IDIST(I,J)
000765	000	C	READ NUMBER OF CONNECTION BETWEEN EACH ELEMENT
000766	000	C	

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000767 000 100 READ(IRD,1002,END=101) IROW,(ICOL(M),IVAL(M)/M,1.5)
000768 000 IF(IROW.EQ.0) GO TO 103
000769 000 IR=IROW
000770 000 103 CONTINUE
000771 000 C CREATE THE MATRIX OF CONNECTIONS
000772 000 C
000773 000 DO 102 M=1,5
000774 000 IF(ICOL(M).EQ.0) GO TO 100
000775 000 IC=ICOL(M)
000776 000 ELTCON(IR,IC)=IVAL(M)
000777 000 ELTCON(IC,IR)=IVAL(M)
000778 000 102 CONTINUE
000779 000 GO TO 100
000780 000 101 IF(LOCOD.EQ.1) GO TO 110
000781 000 C CALCULATE ALL DISTANCES ON THE ASSIGNMENT BOARD AND SUM
000782 000 C THE DISTANCES FROM ONE LOCATION TO ALL OTHER LOCATIONS
000783 000 C
000784 000 XELT=NELT
000785 000 NDIM=50RT(XELT)+1
000786 000 NLOC=NDIM*NDIM
000787 000 I=0
000788 000 DO 106 ILOC=1,NLOC
000789 000 AUX1(ILOC)=0
000790 000 IF(ILOC.LE.I*NDIM) GO TO 104
000791 000 I=I+1
000792 000 J=0
000793 000 104 J=J+1
000794 000 TO=0
000795 000 FROM=ILOC
000796 000 DO 106 M=1,NDIM
000797 000 DO 106 K=1,NDIM
000798 000 TC=TO+1
000799 000 J1=IABS(J-K)
000800 000 I1=IABS(I-M)
000801 000 IF(KODIS.EQ.0) IDIST(FROM,TO)=I1+J1
000802 000 IF(KODIS.EQ.1) IDIST(FROM,TO)=(I1+J1+J1*J1)*0.5
000803 000 106 AUX1(ILOC)=AUX1(ILOC)+IDIST(FROM,TO)
000804 000 GO TO 109
000805 000 C SUM DISTANCES FROM EACH LOCATION (READ EXPLICITLY)
000806 000 C
000807 000 110 DO 105 FROM=1,NLOC
000808 000 AUX1(FROM)=0
000809 000 DO 105 TO=1,NLOC
000810 000 105 AUX1(FROM)=AUX1(FROM)+IDIST(FROM,TO)
000811 000 C SORT SUM OF DISTANCES IN ASCENDING ORDER
000812 000 C
000813 000 109 CALL SORT(AUX1,IDISTG,NLOC,NLOC,1,1)
000814 000 C FIND TOTAL NUMBER OF PIPES AND CONNECTED ELEMENTS
000815 000 C TO THE ELEMENT AND SORT ELEMENTS IN ASCENDING ORDER
000816 000 C
000817 000 DO 107 IELT=1,NELT
000818 000 AUX1(IELT)=0
000819 000 AUX2(IELT)=0
000820 000 DO 107 JELT=1,NELT
000821 000 IF(ELTCON(IELT,JELT).EQ.0) GO TO 107
000822 000 AUX2(IELT)=AUX2(IELT)+1
000823 000 AUX1(IELT)=AUX1(IELT)+ELTCON(IELT,JELT)

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000824 000 107 CONTINUE
000825 000 CALL SORT(AUX1,NWIRES,NELT,NELT,1,1)
000826 000 CALL SORT(AUX2,NCONS,NELT,NELT,1,1)
000827 000 C SUM THE RANKS OF EACH ELEMENT (DESCENDING ORDER)
000828 000 C
000829 000 CALL SORT(NWIRES,AUX1,NELT,NELT,1,1)
000830 000 CALL SORT(NCONS,AUX2,NELT,NELT,1,1)
000831 000 DO 108 IELT=1,NELT
000832 000 108 AUX3(IELT)=AUX1(IELT)+AUX2(IELT)
000833 000 CALL SORT(AUX3,FINRAN,NELT,NELT,-1,1)
000834 000 C FIND THE INITIAL LOCATION OF THE ELEMENTS
000835 000 C ON THE ASSIGNMENT BOARD AND CALCULATE INITIAL BOUND
000836 000 C
000837 000 MELT=0
000838 000 DO 111 IELT=1,NELT
000839 000 ISWCH=1
000840 000 IAS=FINRAN(IELT)
000841 000 IF(ASELT(IAS).EQ.EMPTY) GO TO 115
000842 000 GO TO 111
000843 000 C FIND THE ELEMENT WHICH HAS MOST CONNECTIONS
000844 000 C TO PREVIOUSLY ASSIGNED ELEMENT
000845 000 C
000846 000 113 ISWCH=0
000847 000 C ELEMENTS ARE SORTED IN DESCENDING ORDER ACCORDING
000848 000 C TO MOST CONNECTIONS TO PREVIOUSLY ASSIGNED ELEMENTS
000849 000 C
000850 000 DO 114 J=1,NELT
000851 000 114 AUX1(J)=ELTCON(IAS,J)
000852 000 CALL SORT(AUX1,AUX2,NELT,NELT,-1,1)
000853 000 DO 118 I=1,NELT
000854 000 IF(AUX1(I).EQ.0) GO TO 111
000855 000 IAS=AUX2(I)
000856 000 IF(ASELT(IAS).EQ.EMPTY) GO TO 115
000857 000 118 CONTINUE
000858 000 C MAKE THE ASSIGNMENTS TO NEXT BEST POSITION
000859 000 C AND KEEP THE RECORD OF ASSIGNMENTS
000860 000 115 MELT=MELT+1
000861 000 ILO=IDISTS(MELT)
000862 000 BOARD(ILO)=IAS
000863 000 ASELT(IAS)=ILO
000864 000 ASRANK(MELT)=IAS
000865 000 NASSEL=NASSEL+1
000866 000 ASSEL(NASSEL)=IAS
000867 000 IF(MELT.EQ.NELT) GO TO 120
000868 000 IF(ISWCH.EQ.0) GO TO 111
000869 000 GO TO 113
000870 000 111 CONTINUE
000871 000 120 DO 119 I=1,NLOC
000872 000 119 AUXBOA(I)=BOARD(I)
000873 000 C SORT ELEMENT CONNECTIONS IN DESCENDING ORDER
000874 000 C
000875 000 NE=NELT-1
000876 000 KCON=0
000877 000 DO 116 I=1,NE
000878 000 IB=I+1
000879 000 DO 116 J=IB,NELT
000880 000 KCON=KCON+1

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000881 000 IELTC3(KCON)=ELTCON(I,J)
000882 000 AUX5(KCON)=I
000883 000 116 AUX6(KCON)=J
000884 000 CALL SORT(IELTC3,AUX4,KCON,KCON,-1,1)
000885 000 C MODIFY THE ELEMENT NUMBERS ACCORDING TO SORTING
000886 000 C
000887 000 DO 160 I=1,KCON
000888 000 IELT=AUX4(I)
000889 000 ELEFR(I)=AUX5(IELT)
000890 000 160 ELETO(I)=AUX6(IELT)
000891 000 C SORT DISTANCES BETWEEN LOCATIONS IN ASCENDING ORDER
000892 000 C
000893 000 NE=NLOC-1
000894 000 KDIS=0
000895 000 DO 142 I=1,NE
000896 000 IB=I+1
000897 000 DO 142 J=IB,NLOC
000898 000 KDIS=KDIS+1
000899 000 ILOCC3(KDIS)=IDIST(I,J)
000900 000 AUX5(KDIS)=I
000901 000 142 AUX6(KDIS)=J
000902 000 CALL SORT(ILOCC3,AUX4,KDIS,KDIS,1,1)
000903 000 C MODIFY THE LOCATION NUMBERS ACCORDING TO SORTING
000904 000 C
000905 000 DO 161 I=1,KDIS
000906 000 ILOC=AUX4(I)
000907 000 DISTFR(I)=AUX5(ILOC)
000908 000 161 DISTTO(I)=AUX6(ILOC)
000909 000 C CALCULATE INITIAL CSTAR
000910 000 C
000911 000 DO 121 IELT=1,NELT
000912 000 ILOC=ASELT(IELT)
000913 000 K=IELT+1
000914 000 DO 121 JELT=K,NELT
000915 000 JLOC=ASELT(JELT)
000916 000 121 CSTAR=CSSTAR+ELTCON(IELT,JELT)*IDYST(ILOC,JLOC)
000917 000 C1=CSSTAR
000918 000 CPER=(IPERC*CSSTAR)/100
000919 000 WRITE(IWT,1006) CSTAR
000920 000 DO 122 I=1,NLOC
000921 000 122 WRITE(IWT,1004) I,BOARD(I)
000922 000 ITER=NELT+1
000923 000 C START ITERATIONS ON THE TREE
000924 000 C
000925 000 123 ITER=ITER-1
000926 000 IF(ITER.EQ.0) GO TO 200
000927 000 C DELETE NEXT PREVIOUSLY ASSIGNED ELEMENT
000928 000 C
000929 000 IRE=ASRANK(ITER)
000930 000 ILO=ASELT(IRE)
000931 000 AUXDOA(ILO)=EMPTY
000932 000 ITREE(IRE,ILO)=INFIN
000933 000 ASELT(IRE)=EMPTY
000934 000 NASSEL=NASSEL-1
000935 000 C CALCULATE COST OF DELETED ELEMENT
000936 000 C
000937 000 C UPDATE C1

```



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000938 000 C
000939 000 DO 140 I=1,NASSEL
000940 000 IELT=ASELT(I)
000941 000 ILOC=ASELT(IELT)
000942 000 140 C1=C1-ELTCON(IRE,IELT)*IDIST(ILO,ILOC)
000943 000 C CALCULATE C2 & C3
000944 000 C
000945 000 C2=0
000946 000 C3=0
000947 000 IF(NASSEL.LT.NELT) CALL COST(C2,C3,IC2)
000948 000 CNEW=C1+C2+C3
000949 000 I10=I10+1
000950 000 IF(I10.LE.10) WRITE(IWT,1007) I10,C1,C2,C3
000951 000 IF(CNEW.LT.CPER) GO TO 126
000952 000 124 DO 125 I=1,NLOC
000953 000 125 ITRF(IRE,I)=EMPTY
000954 000 GO TO 123
000955 000 C CHECK IF THERE IS A POSSIBILITY TO ASSIGN THE
000956 000 C REMOVED ELEMENT TO ANY OTHER LOCATION
000957 000 126 DO 130 JLOC=1,NLOC
000958 000 IF(AUXBOA(JLOC).EQ.EMPTY.AND.ITRF(IRE,JLOC).EQ.EMPTY)
000959 000 1 GO TO 127
000960 000 GO TO 130
000961 000 C MAKE THE NEW ASSIGNMENT
000962 000 C
000963 000 127 AUXBOA(JLOC)=IRE
000964 000 ASELT(IRE)=JLOC
000965 000 ITER=ITER+1
000966 000 NASSEL=NASSEL+1
000967 000 C CALCULATE COST OF NEW ASSIGNED ELEMENT
000968 000 C
000969 000 C UPDATE C1
000970 000 C
000971 000 DO 150 I=1,NASSEL
000972 000 IELT=ASELT(I)
000973 000 ILOC=ASELT(IELT)
000974 000 150 C1=C1+ELTCON(IRE,IELT)*IDIST(JLOC,ILOC)
000975 000 C2=0
000976 000 C3=0
000977 000 IF(NASSEL.LT.NELT) CALL COST(C2,C3,IC2)
000978 000 CNEW=C1+C2+C3
000979 000 I10=I10+1
000980 000 IF(I10.LE.10) WRITE(IWT,1007) I10,C1,C2,C3
000981 000 IF(CNEW.GE.CPER) GO TO 123
000982 000 C CHECK IF ALL THE ELEMENTS HAVE BEEN ASSIGNED
000983 000 C
000984 000 IF(ITER.EQ.NELT+1) GO TO 132
000985 000 GO TO 131
000986 000 130 CONTINUE
000987 000 GO TO 124
000988 000 C FIND THE NEXT UNASSIGNED ELEMENT
000989 000 C
000990 000 131 IRE=ASRANK(ITER)
000991 000 GO TO 126
000992 000 C UPDATE ASSIGNMENT BOARD & BOUND
000993 000 C
000994 000 132 CSTAR=CNEW

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000995      000      CPER=(IPERC*STAR)/100
000996      000      DO 133 IJ=1,NLOC
000997      000 133    BOARD(IJ)=AUXBOA(IJ)
000998      000      GO TO 123
000999      000      C          OUTPUT THE RESULTS
001000      000      C
001001      000 200    WRITE(IWT,1005) CSTAR
001002      000      DO 134 I=1,NLOC
001003      000 134    WRITE(IWT,1004) I,BOARD(I)
001004      000      C          FORMAT STATEMENTS
001005      000      C
001006      000 1001   FORMAT(8I5)
001007      000 1002   FORMAT(15,5(15,110))
001008      000 1003   FORMAT(10X,7I10)
001009      000 1004   FORMAT(2X,13,9X,13)
001010      000 1005   FORMAT(///,2X,'OPTIMAL COST:',10,///,1X,'LOCATION:',10,4X,'ELEMENT')
001011      000 1006   FORMAT(///,2X,'INITIAL COST:',10,///,1X,'LOCATION:',10,4X,'ELEMENT')
001012      000 1007   FORMAT(//,2X,'ITERATION:',10,13,' C1:',15,13,' C2:',15,13,' C3:',15)
001013      000 9999   FORMAT( )
001014      000      STOP
001015      000      END
001016      000  DELT,SI TUMER-S*SINAN-BAB,CONT,,,170515041312
001017      000      SUBROUTINE COST(C2,C3,IC2)
001018      000      C
001019      000      C          ROUTINE CALCULATED THE BOARD OF THE GRAPH & BOUND ASSIGNMENT
001020      000      C          PROBLEM AND ALSO CALCULATED THE COST OF INITIAL SOLUTION
001021      000      C
001022      000      C          VARIABLES:
001023      000      C          ASSEL :ASSIGNED ELEMENTS
001024      000      C          C2      :COST OF ASSIGNED ELEMENTS TO UNASSIGNED ONES
001025      000      C          C3      :COST OF ASSIGNED ELEMENTS TO ELEMENTS
001026      000      C          IC2     :=1 IF C2 IS CALCULATED WITH INITIAL C1X
001027      000      C          =0 IF C2 IS CALCULATED BY SORTED ELEMENTS AND LOCATIONS
001028      000      C          NASSEL:NUMBER OF ASSIGNED ELEMENTS
001029      000      C          NUNASE:NUMBER OF UNASSIGNED ELEMENTS
001030      000      C          NUNASL:NUMBER OF UNASSIGNED LOCATIONS
001031      000      C          RHO     :RHO MATRIX USED FOR CALCULATION OF C2
001032      000      C          UNASEL:UNASSIGNED ELEMENTS
001033      000      C          UNASLO:LOCATIONS WHERE ARE UNASSIGNED
001034      000      C
001035      000      C
001036      000      COMMON/AUXILI/AUX5(800),AUX6(600)
001037      000      COMMON/AREA/AUXBOA(40),AUX1(40),AUX2(40),AUX3(40),ITREE(35,40),
001038      000      11DIST(40,40),ELTCOM(55,40),DELT(35),ASSEL(35),DELT,NLOC,NASSEL
001039      000      2,KDIS,KCON,IELTC3(650),J1,LOCN(400),ELEFR(650),ELETO(650),
001040      000      3DISTF(800),DISTTO(600)
001041      000      INTEGER AUXBOA,AUX1,AUX2,AUX3,EMPTY,ELTCOM,ITREE(35,40),UNASEL(35),
001042      000      1UNASLO(40),ASSEL,ASLT,AUX5,AUX6,UNAS,C2,C3
001043      000      2,ELEFR,ELETO,DISTF,DISTTO,FROM,TO
001044      000      C
001045      000      C          FIND ASSIGNED AND UNASSIGNED ELEMENTS & LOCATIONS
001046      000      C
001047      000      IFIN=1
001048      000      IWT=6
001049      000 9999   FORMAT( )
001050      000      EMPTY=0
001051      000      IF(IC2.EQ.0) GO TO 112

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001052 000 NUNASE=0
001053 000 NUNASL=0
001054 000 DO 90 I=1,NELT
001055 000 IF (ASELT(I).EQ.EMPTY) GO TO 91
001056 000 GO TO 90
001057 000 C FIND UNASSIGNED ELEMENTS
001058 000 C
001059 000 91 NUNASE=NUNASE+1
001060 000 UNASEL(NUNASE)=1
001061 000 90 CONTINUE
001062 000 C FIND UNASSIGNED LOCATIONS
001063 000 C
001064 000 DO 92 I=1,NLOC
001065 000 IF (AUXBOA(I).NE.EMPTY) GO TO 92
001066 000 NUNASL=NUNASL+1
001067 000 UNASLO(NUNASL)=1
001068 000 92 CONTINUE
001069 000 IF (NASSEL.EQ.0) GO TO 110
001070 000 C CALCULATE C2
001071 000 C
001072 000 C CREATE RHO MATRIX
001073 000 C
001074 000 101 DO 102 I=1,NUNASE
001075 000 M=UNASEL(I)
001076 000 DO 102 J=1,NUNASL
001077 000 L=UNASLO(J)
001078 000 C CHECK IF WE ARE ALLOWED TO ASSIGN ELEMENT M TO LOCATION L
001079 000 C
001080 000 IF (ITREE(M,L).EQ.EMPTY) GO TO 99
001081 000 RHO(I,J)=ITREE(M,L)
001082 000 GO TO 102
001083 000 99 RHO(I,J)=0
001084 000 DO 98 K=1,NASSEL
001085 000 IE=ASSEL(K)
001086 000 IL=ASELT(IE)
001087 000 RHO(I,J)=RHO(I,J)+ELTCON(M,IE)*IDIST(L,IL)
001088 000 98 CONTINUE
001089 000 102 CONTINUE
001090 000 C FIND MINIMUMS OF ROWS (STORED IN AUX1)
001091 000 C
001092 000 DO 103 I=1,NUNASE
001093 000 AUX1(I)=RHO(I,1)
001094 000 DO 103 J=2,NUNASL
001095 000 IF (AUX1(I).LE.RHO(I,J)) GO TO 103
001096 000 AUX1(I)=RHO(I,J)
001097 000 103 CONTINUE
001098 000 DO 105 I=1,NUNASE
001099 000 DO 105 J=1,NUNASL
001100 000 105 RHO(I,J)=RHO(I,J)-AUX1(I)
001101 000 C FIND MINIMUMS OF COLUMNS (STORED IN AUX2)
001102 000 C
001103 000 DO 104 J=1,NUNASL
001104 000 AUX2(J)=RHO(1,J)
001105 000 DO 104 I=2,NUNASE
001106 000 IF (AUX2(J).LE.RHO(I,J)) GO TO 104
001107 000 AUX2(J)=RHO(I,J)
001108 000 104 CONTINUE

```

001109	000	C	ADD MINIMUMS OF ROWS AND COLUMNS
001110	000	C	
001111	000		DO 106 I=1,NUNASE
001112	000	106	C2=C2+AUX1(I)+AUX2(I)
001113	000		IF(NUNASE.LE.1) RETURN
001114	000		GO TO 110
001115	000	C	CALCULATE C2 BY SORTING METHOD
001116	000	C	
001117	000	112	K=0
001118	000		IF(NASSEL.EQ.0) GO TO 110
001119	000		DO 116 I=1,KCON
001120	000	C	CHECK WHETHER EITHER ONE OF THE ELEMENTS ASSIGNED
001121	000	C	
001122	000		FROM=ELEFR(I)
001123	000		TO=ELETO(I)
001124	000		IF(ASELT(FROM).EQ.EMPTY) GO TO 113
001125	000		IF(ASELT(TO).NE.EMPTY) GO TO 116
001126	000		GO TO 114
001127	000	113	IF(ASELT(TO).EQ.EMPTY) GO TO 116
001128	000		K=K+1
001129	000		AUX6(K)=FROM
001130	000		GO TO 115
001131	000	114	K=K+1
001132	000		AUX6(K)=TO
001133	000	115	AUX5(K)=IELTC3(I)
001134	000	116	CONTINUE
001135	000	C	CHECK WHETHER EITHER ONE OF THE LOCATIONS ASSIGNED
001136	000	C	
001137	000		M=0
001138	000		DO 130 I=1,KDIS
001139	000		FROM=DISFR(I)
001140	000		TO=DISTO(I)
001141	000		IF(AUXBOA(FROM).EQ.EMPTY) GO TO 121
001142	000		IF(AUXBOA(TO).NE.EMPTY) GO TO 130
001143	000		GO TO 123
001144	000	121	IF(ASELT(TO).EQ.EMPTY) GO TO 130
001145	000		MEM=1
001146	000		IT=FROM
001147	000		GO TO 124
001148	000	123	MEM=1
001149	000		IT=TO
001150	000	124	IJ=AUX6(M)
001151	000		IF(ITREE(IJ,IT).NE.EMPTY) GO TO 130
001152	000		C2=C2+AUX5(M)*ILOCC3(I)
001153	000		IF(M.EQ.K) GO TO 110
001154	000	130	CONTINUE
001155	000	C	CALCULATE C3
001156	000	C	
001157	000	110	K=0
001158	000		DO 135 I=1,KCON
001159	000	C	CHECK WHETHER BOTH THE ELEMENTS ARE ASSIGNED
001160	000	C	
001161	000		FROM=ELEFR(I)
001162	000		TO=ELETO(I)
001163	000		IF(ASELT(FROM).EQ.EMPTY.AND.ASELT(TO).EQ.EMPTY) GO TO 136
001164	000		GO TO 135
001165	000	136	K=K+1

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001166 000 AUX5(K)=IFLTC3(I)
001167 000 135 CONTINUE
001168 000 C CHECK WHETHER BOTH THE LOCATIONS ARE UNASSIGNED
001169 000 C
001170 000 M=0
001171 000 DO 139 I=1,KDIS
001172 000 FROM=DISTFR(I)
001173 000 TO=DISTTO(I)
001174 000 IF(AUXBOA(FROM).EQ.EMPTY.AND.AUXBOA(TO).EQ.EMPTY) GO TO 141
001175 000 GO TO 139
001176 000 141 M=M+1
001177 000 C3=C3+AUX5(M)*ILOCC3(I)
001178 000 IF(M.EQ.K) GO TO 150
001179 000 139 CONTINUE
001180 000 150 RETURN
001181 000 END
001182 000 SUBROUTINE SORT(WORK,IWORK,NELT,KELT,KOD1,INXSRT)
001183 000 C
001184 000 C SORTING ROUTINE OF AN INTEGER VECTOR
001185 000 C PROGRAM SORTS THE ELEMENTS OF A VECTOR AND INDEXES
001186 000 C ACCORDING TO ASCENDING OR DESCENDING ORDER
001187 000 C
001188 000 C VARIABLES:
001189 000 C WORK :VECTOR TO BE SORTED
001190 000 C IWORK :SORTED INDEXES
001191 000 C WAIT :TEMPORARY AREA
001192 000 C KELT :NUMBER OF ELEMENTS IN THE VECTOR IWORK
001193 000 C NELT :NUMBER OF ELEMENTS IN THE VECTOR WORK
001194 000 C KOD1 :=1 , VECTOR IS TO BE SORTED IN ASCENDING ORDER
001195 000 C :=-1 , VECTOR IS TO BE SORTED IN DESCENDING ORDER
001196 000 C SMALL :SMALLEST ELEMENT IN THE VECTOR AT THE MOMENT OF SORTING
001197 000 C I :ITERATION NUMBER
001198 000 C INXSRT:=1 , INDEX SORTING IS DESIRED
001199 000 C :0 OTHERWISE
001200 000 C J :ELEMENT NUMBER
001201 000 C
001202 000 C
001203 000 INTEGER WORK,WAIT,SMALL
001204 000 DIMENSION WORK(NELT),IWORK(KELT)
001205 000 IWT=6
001206 000 C WHEN INDEX SORTING IS DESIRED NELT MUST BE EQUAL TO KELT
001207 000 C
001208 000 IF(INXSRT.EQ.0) GO TO 96
001209 000 IF(NELT.EQ.KELT) GO TO 94
001210 000 WRITE(IWT,1002) INXSRT
001211 000 1002 FORMAT(/// STOP: CHECK THE DIMENSIONS OF ARRAYS TO BE SORTED'
001212 000 1,///' INXSRT=',I6)
001213 000 STOP
001214 000 C INITIALIZE INDEX ARRAY
001215 000 C
001216 000 94 DO 95 IELT=1,NELT
001217 000 95 IWORK(IELT)=IELT
001218 000 96 I=1
001219 000 IF(INXSRT.EQ.0) INS=1
001220 000 IF(INXSRT.EQ.1) INS=2
001221 000 NELT1=NELT+1
001222 000 100 J=I

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001223 000 C      INITIALIZE THE SMALLEST VALUE OF NEXT ITERATION ARBITRARILY
001224 000 C
001225 000      SMALL=WORK(I)
001226 000 102    J=J+1
001227 000      IF(J.EQ.NFLT1) GO TO 101
001228 000 C      CHECK WHETHER NEXT ELEMENT IS LESS THAN THE
001229 000 C      PRESENT SMALL VALUE
001230 000 C
001231 000      IF(SMALL.LE.WORK(J)) GO TO 102
001232 000 C      CHANGE THE ELEMENTS ACCORDING TO ASCENDING ORDER
001233 000 C
001234 000      SMALL=WORK(J)
001235 000      WAIT=WORK(I)
001236 000      WORK(I)=WORK(J)
001237 000      WORK(J)=WAIT
001238 000 C      CHANGE INDEXES TO ASCENDING ORDER
001239 000 C
001240 000      GO TO (102,103),INS
001241 000 103    WAIT=IWORK(I)
001242 000      IWORK(I)=IWORK(J)
001243 000      IWORK(J)=WAIT
001244 000      GO TO 102
001245 000 C      NEXT ITERATION
001246 000 C
001247 000 101    I=I+1
001248 000 C      END OF ITERATION
001249 000 C
001250 000      IF(I.LT.NFLT1) GO TO 100
001251 000 C      CHECK IF DESCENDING ORDER IS DESIRED
001252 000 C
001253 000      IF(KOD1.EQ.1) RETURN
001254 000      IF(KOD1.EQ.-1) GO TO 106
001255 000      WRITE(IWT,1001)
001256 000 1001   FORMAT(/,5X,'STOP KOD1 IS NOT PROPER')
001257 000      STOP
001258 000 C      REARRANGE THE ELEMENTS AND INDEXES IN DESCENDING ORDER
001259 000 C
001260 000 106    NB=1
001261 000      NE=NFLT
001262 000      N=NFLT/2
001263 000      DO 107 M=1,N
001264 000      WAIT=WORK(NE)
001265 000      WORK(NE)=WORK(NB)
001266 000      WORK(NB)=WAIT
001267 000      GO TO (105,104),INS
001268 000 104    WAIT=IWORK(NE)
001269 000      IWORK(NE)=IWORK(NB)
001270 000      IWORK(NB)=WAIT
001271 000 105    NB=NB+1
001272 000 107    NE=NE-1
001273 000      RETURN
001274 000      END
001275 000      QFOR,IS MAIN
001276 000      DELT,SI TUMER-S*SINAN-QAP5.MAY10,,143216053312
001277 000      COMMON IC1,IC2,IC3,MFCIL,NLOC,N2,M2,LPCST,ITOT
001278 000      COMMON MSTP(04000),IWORK(2000),KSNSW(60)
001279 000      COMMON KLBND,KUPDT

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QAP3 1
QAP3 2
QAP3 3

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001280	000	INTEGER F(20,20),D(20,20),DISTI(400),DISTJ(400),INTRI(400)	QAP3	4
001281	000	INTEGER INTRJ(400),LEVIL(10),LOCAT(20),FCLTY(20),PTCST(20)	QAP3	5
001282	000	INTEGER FIXED(20),WORK1(20,20),CNCST(20),LIMED(20)	QAP3	6
001283	000	INTEGER IPASS(20),ICOST(100),ILOCT(100,20),FOUND(100)	QAP3	7
001284	000	INTEGER FIX(20),WIDPF(20)	QAP3	8
001285	000	LOGICAL FCFLG(20),FLOGT(20),FLGDS(400),FLGIN(400),BANED(20,20)	QAP3	9
001286	000		QAP3	10
001287	000	LOGICAL WORK2(1300),WORK4(1300)	QAP3	11
001288	000	INTEGER WORK3(0120),WORK7(0120),WORK6(0020)	QAP3	12
001289	000		QAP3	13
001290	000		QAP3	14
001291	000	EQUIVALENCE( MSTER(0001),F(1,1))	QAP3	15
001292	000	EQUIVALENCE( MSTER(0001),D(1,1))	QAP3	16
001293	000	EQUIVALENCE( MSTER(0001),DISTI(1))	QAP3	17
001294	000	EQUIVALENCE( MSTER(1201),INTRI(1))	QAP3	18
001295	000	EQUIVALENCE( MSTER(1201),INTRJ(1))	QAP3	19
001296	000	EQUIVALENCE( MSTER(2001),LEVIL(1))	QAP3	20
001297	000	EQUIVALENCE( MSTER(2001),LOCAT(1))	QAP3	21
001298	000	EQUIVALENCE( MSTER(2001),FCLTY(1))	QAP3	22
001299	000	EQUIVALENCE( MSTER(2001),PTCST(1))	QAP3	23
001300	000	EQUIVALENCE( MSTER(2001),FIXED(1))	QAP3	24
001301	000	EQUIVALENCE( MSTER(2001),ILOCT(1))	QAP3	25
001302	000	EQUIVALENCE( MSTER(2001),FOUND(1))	QAP3	26
001303	000	EQUIVALENCE( MSTER(2001),CNCST(1))	QAP3	27
001304	000	EQUIVALENCE( MSTER(2001),LIMED(1))	QAP3	28
001305	000	EQUIVALENCE( MSTER(2001),IPASS(1))	QAP3	29
001306	000	EQUIVALENCE( MSTER(3001),ICOST(1,1))	QAP3	30
001307	000	EQUIVALENCE( MSTER(3001),ILOCT(1,1))	QAP3	31
001308	000	EQUIVALENCE( MSTER(3001),FOUND(1,1))	QAP3	32
001309	000	EQUIVALENCE( MSTER(3001),FIX(1))	QAP3	33
001310	000		QAP3	34
001311	000	EQUIVALENCE( IWORK(0001),WORK1(1,1))	QAP3	35
001312	000		QAP3	36
001313	000	EQUIVALENCE( MSTER(2401),WORK7(1))	QAP3	37
001314	000	EQUIVALENCE( MSTER(2521),WORK8(1))	QAP3	38
001315	000		QAP3	39
001316	000		QAP3	40
001317	000		QAP3	41
001318	000	READ THE KEYS SETTINGS FOR THE MAIN PROGRAM	QAP3	42
001319	000	READ(5,3)(KSNSW(I),I=01,20)	QAP3	43
001320	000	READ THE KEY SETTINGS FOR SUBROUTINE LWBND	QAP3	44
001321	000	READ(5,3)(KSNSW(I),I=21,30)	QAP3	45
001322	000	READ THE PROBLEM DATA	QAP3	46
001323	000	10 CALL DTAIN(JOB,IFRAC,NFIXD,ISTOP,MORE,ITERM)	QAP3	47
001324	000		QAP3	48
001325	000	PRINT THE INITIAL DATA IF YOU WISH	QAP3	49
001326	000	IF(KSNSW(01).EQ.0)GO TO 110	QAP3	50
001327	000	WRITE(6,4)JOB,NFCIL,NLOCT	QAP3	51
001328	000	DO 70 I=1,2	QAP3	52
001329	000	70 WRITE(6,2)	QAP3	53
001330	000	WRITE(6,7)	QAP3	54
001331	000	WRITE(6,5)	QAP3	55
001332	000	DO 80 I=1,NLOCT	QAP3	56
001333	000	80 WRITE(6,01)(D(I,J),J=1,NLOCT)	QAP3	57
001334	000	WRITE(6,8)	QAP3	58
001335	000	DO 90 I=1,NFCIL	QAP3	59
001336	000	90 WRITE(6,01)(F(I,J),J=1,NFCIL)	QAP3	60

001337	000	WRITE(6,7)	GAP3	61
001338	000	WRITE(6,51)LBCST,1TOT	GAP3	62
001339	000	C INITIALIZE THE COST	GAP3	63
001340	000	110 IC1=0	GAP3	64
001341	000	N2=NLOCT*NLOCT	GAP3	65
001342	000	M2=NFCIL*NFCIL	GAP3	66
001343	000	IF(NFIXD.EQ.1)GO TO 125	GAP3	67
001344	000	DO 120 I=1,10	GAP3	68
001345	000	120 WRITE(6,98)NFXD	GAP3	69
001346	000	GO TO 900	GAP3	70
001347	000	C UNFLAG ALL THE LOCATIONS AND THE FACILITIES	GAP3	71
001348	000	125 DO 130 I=1,NLOCT	GAP3	72
001349	000	130 FLOGT(I)=.FALSE.	GAP3	73
001350	000	DO 140 J=1,NFCIL	GAP3	74
001351	000	140 FCFLG(J)=.FALSE.	GAP3	75
001352	000	C	GAP3	76
001353	000	DO 150 I=1,NFCIL	GAP3	77
001354	000	IPASS(I)=0	GAP3	78
001355	000	DO 150 J=1,NLOCT	GAP3	79
001356	000	150 BAND(I,J)=.FALSE.	GAP3	80
001357	000	C RANK THE DISTANCES IN AN ASCENDING ORDER	GAP3	81
001358	000	CALL ARRNG(N2,1,2)	GAP3	82
001359	000	C PRINT THE MATRIX OF ARRANGED DISTANCES	GAP3	83
001360	000	IF(KSNSW(02).EQ.0)GO TO 200	GAP3	84
001361	000	WRITE(6,21)	GAP3	85
001362	000	WRITE(6,01)(DISTI(I),I=1,M2)	GAP3	86
001363	000	WRITE(6,07)	GAP3	87
001364	000	WRITE(6,01)(DISTJ(I),I=1,M2)	GAP3	88
001365	000	C	GAP3	89
001366	000	C FLAG ALL THE DISTANCES	GAP3	90
001367	000	200 DO 210 I=1,N2	GAP3	91
001368	000	210 FLGDS(I)=.FALSE.	GAP3	92
001369	000	C	GAP3	93
001370	000	C RANK THE DINTERACTIONS IN A DESCENDING ORDER	GAP3	94
001371	000	CALL ARRNG(M2,2,2)	GAP3	95
001372	000	C PRINT THE INTERACTIONS IF YOU WISH	GAP3	96
001373	000	IF(KSNSW(03).EQ.0)GO TO 310	GAP3	97
001374	000	WRITE(6,22)	GAP3	98
001375	000	WRITE(6,01)(INTRI(I),I=1,M2)	GAP3	99
001376	000	WRITE(6,07)	GAP3	100
001377	000	WRITE(6,01)(INTRJ(I),I=1,M2)	GAP3	101
001378	000	C	GAP3	102
001379	000	C FLAGG ALL THE INTERACTIONS	GAP3	103
001380	000	310 DO 320 I=1,M2	GAP3	104
001381	000	320 FLGIN(I)=.FALSE.	GAP3	105
001382	000	C INITIALIZE WORK7	GAP3	106
001383	000	DO 325 I=1,120	GAP3	107
001384	000	325 WORK7(I)=0	GAP3	108
001385	000	C INITIALIZE WORK8	GAP3	109
001386	000	DO 330 I=1,1300	GAP3	110
001387	000	330 WORK8(I)=.FALSE.	GAP3	111
001388	000	DO 335 I=1,20	GAP3	112
001389	000	335 CMCST(I)=0	GAP3	113
001390	000	C	GAP3	114
001391	000	C	GAP3	115
001392	000	TIME=0	GAP3	116
001393	000	IEVAL=0	GAP3	117



001394	000	IUBCS=0	QAP3 118
001395	000	KOMPL=0	QAP3 119
001396	000	FRAC=IFRAC	QAP3 120
001397	000	FRAC=0.01*FRAC	QAP3 121
001398	000	C	QAP3 122
001399	000	C INITIALIZE COUNTERS	QAP3 123
001400	000	KLBN0=0	QAP3 124
001401	000	KUPDT=0	QAP3 125
001402	000	KFATH=0	QAP3 126
001403	000	KTERM=0	
001404	000	KOMIT=100*NFCIL	
001405	000	C	QAP3 127
001406	000	C CALCULATE A STARTING LOWER BOUND	QAP3 128
001407	000	IF(LBCST.GT.0)GO TO 340	QAP3 129
001408	000	CALL LWBND(0,0,0,0)	QAP3 130
001409	000	340 LBTC=LBCST	QAP3 131
001410	000	IF(KSNSW(04).EQ.1)WRITE(6,17)LPIC	QAP3 132
001411	000	C	QAP3 133
001412	000	C PROCEED TO OBTAIN A FIRST FEASIBLE SOLUTION ALONG THE RIGHTMOST	QAP3 134
001413	000	C BRANCH OR THE TREE	QAP3 135
001414	000	C A) MAKE THW ASSIGNMENT OF THE FIRST ELEMENT TO THE FIRST LOCATION	QAP3 136
001415	000	ITER=1	QAP3 137
001416	000	LEVY=1	QAP3 138
001417	000	L2= FIX(1)	QAP3 139
001418	000	J=WHERE(1)	QAP3 140
001419	000	LEVEL(L2)=1	QAP3 141
001420	000	FIXED(1)=L2	QAP3 142
001421	000	LOCAT(L2)=J	QAP3 143
001422	000	LIXED(1)=J	QAP3 144
001423	000	PTCST(1)=0	QAP3 145
001424	000	CMCST(1)=0	QAP3 146
001425	000	FCLTY(1)=L2	QAP3 147
001426	000	FCFLG(L2)=.TRUE.	QAP3 148
001427	000	FLOGT(J)=.TRUE.	QAP3 149
001428	000	L1=J	QAP3 150
001429	000	IF(KSNSW(05).EQ.1)WRITE(6,9)ITER,LEVY,L2,L1	QAP3 151
001430	000	CALL UPDATE(L2,L1,2)	QAP3 152
001431	000	C B) CHOOSE THE ELEMENT WITH MAXIMUM INTERACTIONS WITH THE ONE ASSIGNED	QAP3 153
001432	000		
001433	000		
001434	000	C AT THE PRECEDING LEVEL	QAP3 154
001435	000	+ 43? LSVY=LSVY+A++ + + +	QAP3 155
001436	000	IPASS(LEVY)=IPASS(LEVY)+1	QAP3 156
001437	000		
001438	000	ITER=ITER+1	QAP3 157
001439	000	JET=FCLTY(LEVY-1)	QAP3 158
001440	000	ISAV=-32767	QAP3 159
001441	000	DO 460 I=1,NFCIL	QAP3 160
001442	000	IF(FCFLG(I))GO TO 460	QAP3 161
001443	000	IF(F(I,JET).LE.ISAV.AND.F(JET,I).LE.ISAV)GO TO 460	QAP3 162
001444	000	ISAV=F(I,JET)	QAP3 163
001445	000	IF(F(JET,I).GT.F(I,JET))ISAV=F(JET,I)	QAP3 164
001446	000	KEEP=I	QAP3 165
001447	000	460 CONTINUE	QAP3 166
001448	000	470 FCFLG(KEEP)=.TRUE.	QAP3 167
001449	000	LEVEL(KEEP)=LEVY	QAP3 168
001450	000	C C) ASSIGN THE ELEMENT TO A LOCATION SO THAT THE COST OF SUCH AN ASSI	QAP3 169

001451	000	C	IS MINIMAL	0AP3 170
001452	000	510	IWD=32767	0AP3 171
001453	000		DO 530 J=1,NLOCT	0AP3 172
001454	000		IF(DANED(KEEP,J))GO TO 550	0AP3 173
001455	000		IF(FLOGT(J))GO TO 530	0AP3 174
001456	000		KAM=0	0AP3 175
001457	000		DO 520 I=1,NFCIL	0AP3 176
001458	000		IF(I.EQ.KEEP)GO TO 520	0AP3 177
001459	000		IF(.NOT.FCFLG(I))GO TO 520	0AP3 178
001460	000		L=LOCAT(I)	0AP3 179
001461	000		ISAVV=F(KEEP,I)	0AP3 180
001462	000		IF(F(I,KEEP).GT.F(KEEP,I))ISAVV=F(I,KEEP)	0AP3 181
001463	000		KAM=KAM+(ISAVV*(J-L))	0AP3 182
001464	000	520	CONTINUE	0AP3 183
001465	000		IF(KAM.GT.IWD)GO TO 530	0AP3 184
001466	000		IWD=KAM	0AP3 185
001467	000		LOCAT(KEEP)=J	0AP3 186
001468	000	530	CONTINUE	0AP3 187
001469	000		IF(IWD.EQ.32767)GO TO 760	0AP3 188
001470	000		IF(KSNSW(05).EQ.1)WRITE(6,9)ITER,LEVY,KEEP,LOCAT(KEEP),IWD	0AP3 189
001471	000		J=LOCAT(KEEP)	0AP3 190
001472	000		FCLTY(LEVY)=KEEP	0AP3 191
001473	000		FLOGT(J)=.TRUE.	0AP3 192
001474	000		IUBCS=IUBCS+IWD	0AP3 193
001475	000		FIXED(LEVY)=KEEP	0AP3 194
001476	000		PTCST(LEVY)=IWD	0AP3 195
001477	000		CALL UPDATE(KEEP,J,J)	0AP3 196
001478	000		IF(KSNSW(06).EQ.0)GO TO 600	0AP3 197
001479	000		CMCST(LEVY)=IUBCS	0AP3 198
001480	000		LIXED(LEVY)=J	0AP3 199
001481	000	C		0AP3 200
001482	000	C	CHECK WHETHER WE HAVE REACHED A TERMINAL NODE OR NOT	0AP3 201
001483	000	580	IF(LEVY.LT.NFCIL.AND.IUBCS.GT.0.0)GO TO 400	0AP3 202
001484	000		IF(LEVY.EQ.NFCIL)GO TO 400	0AP3 203
001485	000	C	CHECK WHETHER IT PAYS TO ENTER THE SEARCH ALONG THIS BRANCH	0AP3 204
001486	000		CALL LBND(LEVY,ITER,END=1)	0AP3 205
001487	000		IF(ILDCST.LT.ITOT)GO TO 600	0AP3 206
001488	000		KFATH=KFATH+1	0AP3 207
001489	000		GO TO 620	0AP3 208
001490	000	505	LEVY=LEVY+1	0AP3 209
001491	000		IPASS(LEVY)=IPASS(LEVY)+1	0AP3 210
001492	000		GO TO 730	0AP3 211
001493	000	C	E) NOW WE HAVE A COMPLETE SOLUTION , PRINT IT IF YOU WISH	0AP3 212
001494	000	590	KOMPL=1	0AP3 213
001495	000		IC1=IUBCS	0AP3 214
001496	000		IF(IUBCS.GE.ITOT)GO TO 620	0AP3 215
001497	000		ITOT=IUBCS	0AP3 216
001498	000		IEVAL=IEVAL+1	0AP3 217
001499	000		ICOST(IEVAL)=IUBCS	0AP3 218
001500	000		FOUND(IEVAL)=ITER	0AP3 219
001501	000		DO 595 I=1,NFCIL	0AP3 220
001502	000		ILOCT(IEVAL,I)=LOCAT(I)	0AP3 221
001503	000	595	CONTINUE	0AP3 222
001504	000	600	IF(KSNSW(06).EQ.0)GO TO 610	0AP3 223
001505	000		WRITE(6,11)ITER	0AP3 224
001506	000		WRITE(6,12)	0AP3 225
001507	000		DO 605 I=1,LEVY	0AP3 226

001508	000	605 WRITE(6,14)I, FIXED(I), LIXED(Y), PTCST(I), CMCST(I)	QAP3 227
001509	000	CALL TIMEOUT	
001510	000	C	QAP3 228
001511	000	C	QAP3 229
001512	000	C	QAP3 230
001513	000	C REDUCE THE CURRENT LOWER BOUND	QAP3 231
001514	000	610 DIF=ITOT-LBTC	QAP3 232
001515	000	KETOT=ITOT	QAP3 233
001516	000	IF(ITOT.LE.LBTC)GO TO 800	
001517	000	KELBC=LBTC	QAP3 234
001518	000	IF(DIF.LE.ISTOP)KTERM=100	
001519	000	IF(DIF.LE.ISTOP)GO TO 611	QAP3 235
001520	000	DIF=(FRAC*DIF)	QAP3 236
001521	000	ITOT=LBTC+DIF	QAP3 237
001522	000	611 IF(KSNSW(08).EQ.1)WRITE(6,36)KELPC, ITOT, KETOT	QAP3 238
001523	000	C	QAP3 239
001524	000	C F)STORE THE CONFIGURATION OF THE CURRENT NODE SO THAT WE CAN RESTORE	QAP3 240
001525	000	IT IF WE FAIL TO OBTAIN A BETTER SOLUTION UNDER THE CURRENT CEILING	QAP3 241
001526	000	DO 612 I=1,1300	QAP3 242
001527	000	WORK2(I)=.FALSE.	QAP3 243
001528	000	IF(WORK8(I))WORK2(I)=.TRUE.	QAP3 244
001529	000	612 CONTINUE	QAP3 245
001530	000	DO 615 I=1,120	QAP3 246
001531	000	615 WORK3(I)=WORK7(I)	QAP3 247
001532	000	DO 616 I=1,20	QAP3 248
001533	000	616 WORK6(I)=CMCST(I)	QAP3 249
001534	000	C	QAP3 250
001535	000	C *** BACKTRACK TO THE NEXT LEVEL UP. THE TREE	QAP3 251
001536	000	C A) FREE THE FACILITY AT THAT LEVEL	QAP3 252
001537	000	620 DO 640 I=1,NFCIL	QAP3 253
001538	000	IF(LEVEL(I).EQ.LEVY)GO TO 640	QAP3 254
001539	000	640 CONTINUE	QAP3 255
001540	000	C	QAP3 256
001541	000	650 FCFLG(I)=.FALSE.	QAP3 257
001542	000	DO 670 J=1,NLOCT	QAP3 258
001543	000	IF(J.EQ.LOCAT(I))GO TO 680	QAP3 259
001544	000	670 CONTINUE	QAP3 260
001545	000	680 FLOGT(J)=.FALSE.	QAP3 261
001546	000	IWD=-PTCST(LEVY)	QAP3 262
001547	000	IUBCS=IUBCS+IWD	QAP3 263
001548	000	BANED(I,J)=.TRUE.	QAP3 264
001549	000	C	QAP3 265
001550	000	C	QAP3 266
001551	000	C B) UPDATE THE FLAPS OF THE TWO ORDERED ARRAYS	QAP3 267
001552	000	CALL UPDATE(I,J,1)	QAP3 268
001553	000	C	QAP3 269
001554	000	C C) CALCULATE A LB ON ALL SOLUTIONS EXCLUDING THE LAST	QAP3 270
001555	000	CALL LWBND(LEVY,ITER,IWD,2)	QAP3 271
001556	000	C	QAP3 272
001557	000	C D) CHECK WHETHER IT PAYS TO RESUME THE SEARCH ALONG THIS BRANCH	QAP3 273
001558	000	IF(LDCST.LT.ITOT)GO TO 730	QAP3 274
001559	000	KFATH=KFATH+1	QAP3 275
001560	000	KEEP=I	QAP3 276
001561	000	GO TO 740	QAP3 277
001562	000	C	QAP3 278
001563	000	C RETRIEVE THE INDEX OF THE FACILITY AT THAT LEVEL	QAP3 279
001564	000	730 KEEP=FIXED(LEVY)	QAP3 280

001565	000	ITER=ITER+1	QAP3 281
001566	000	GO TO 470	QAP3 282
001567	000	C	QAP3 283
001568	000	740 DO 760 J=1,NLOCT	QAP3 284
001569	000	760 RANED(KEEP,J)=.FALSE.	QAP3 285
001570	000	FCFLG(KEEP)=.FALSE.	QAP3 286
001571	000	LEVY=LEVY-1	QAP3 287
001572	000	ITER=ITER+1	QAP3 288
001573	000	IF(ITER.LT.KOMIT)GO TO 764	
001574	000	KOMIT=KOMIT+(100*NFCIL)	
001575	000	KEEP=0	
001576	000	DO 762 I=1,NFCIL	
001577	000	762 KEEP=KEEP+IPASS(I)	
001578	000	WRITE(6,52)ITER	
001579	000	WRITE(6,28)(IPASS(I),I=1,NFCIL)	
001580	000	WRITE(6,29)KEEP	
001581	000	CALL TIMEOUT	
001582	000	764 IF(LEVY.NE.0.AND.ITER.LT.ITERM)GO TO 620	QAP3 289
001583	000	IF(LEVY.E-.0.AND.ITER.LT.ITERM)GO TO 775	QAP3 290
001584	000	WRITE(6,6)	QAP3 291
001585	000	WRITE(6,39)LBTC,ITOT	QAP3 292
001586	000	WRITE(6,7)	QAP3 293
001587	000	WRITE(6,52)ITER	
001588	000	CALL TIMEOUT	
001589	000	DO 765 I=1,2	QAP3 294
001590	000	765 WRITE(6,41)ITER	QAP3 295
001591	000	GO TO 810	QAP3 296
001592	000	C	QAP3 297
001593	000	C SINCE NO BETTER SOLUTION WAS FOUND , FLIP THE BOARDS , RESTORE THE	QAP3 298
001594	000	C ARRAYS AND RETURN TO 620	QAP3 299
001595	000	775 IF(KNSW(08).EQ.1)WRITE(6,37)LBTC,ITOT	QAP3 300
001596	000	IF(KNSW(08).EQ.1)WRITE(6,37)ITER	QAP3 301
001597	000	IF(KTERM.EQ.100)GO TO 800	
001598	000	CALL TIMEOUT	
001599	000	LBTC=ITOT	QAP3 302
001600	000	IF(MORE.EQ.0)GO TO 800	
001601	000	KELUC=LBTC	QAP3 303
001602	000	IC1=KETOT	QAP3 304
001603	000	IUBCS=KETOT	QAP3 305
001604	000	ITOT=KETOT	QAP3 306
001605	000	DIF=ITOT-LBTC	QAP3 307
001606	000	IF(DIF.LT.ISTOP)GO TO 778	QAP3 308
001607	000	DIF=FRAC*DIF	QAP3 309
001608	000	GO TO 780	
001609	000	778 IF(KNSW(08).EQ.1)WRITE(6,37)DIF	QAP3 310
001610	000	IF(KTERM.EQ.1)GO TO 800	
001611	000	KTERM=1	
001612	000	ITOT=KETOT	
001613	000	GO TO 782	
001614	000	C	QAP3 312
001615	000	780 KETOT=ITOT	QAP3 313
001616	000	ITOT=LBTC+DIF	QAP3 314
001617	000	782 DO 785 I=1,1300	QAP3 315
001618	000	WORK8(I)=.FALSE.	QAP3 316
001619	000	IF(WORK2(I))WORK8(I)=.TRUE.	QAP3 317
001620	000	785 CONTINUE	QAP3 318
001621	000	DO 790 I=1,120	QAP3 319

001622	000	790	WORK7(I)=WORK3(I)			QAP3	320
001623	000		DO 795 I=1,20			QAP3	321
001624	000	795	CMCST(I)=WORK6(I)			QAP3	322
001625	000		LEVY=NFCIL			QAP3	323
001626	000	C				QAP3	324
001627	000		IF(KNSW(08).EQ.1)WRITE(6,38)LRTC,ITOT			QAP3	325
001628	000		GO TO 620			QAP3	326
001629	000	C				QAP3	327
001630	000	800	WRITE(6,16)			QAP3	328
001631	000	810	KEEP=0			QAP3	329
001632	000		FTHRIO=100.*FLOAT(KFATH)/FLOAT(KLBND)			QAP3	330
001633	000		CALL TIMOUT				
001634	000		DO 820 I=1,NFCIL			QAP3	331
001635	000		LIXED(I)=I			QAP3	332
001636	000	820	KEEP=KEEP+IPASS(I)			QAP3	333
001637	000		WRITE(6,23)			QAP3	334
001638	000		WRITE(6,24)NFCIL			QAP3	335
001639	000		WRITE(6,26)NLOCT			QAP3	336
001640	000		WRITE(6,07)			QAP3	337
001641	000		WRITE(6,47)NFI XD			QAP3	338
001642	000		WRITE(6,48)			QAP3	339
001643	000		DO 825 I=1,NFI XD			QAP3	340
001644	000	825	WRITE(6,49)I,FI X(I),WHERE(I)			QAP3	341
001645	000		WRITE(6,08)			QAP3	342
001646	000		DO 830 I=1,NFCIL			QAP3	343
001647	000	830	WRITE(6,1)(F(I,J),J=1,NFCIL)			QAP3	344
001648	000		WRITE(6,05)			QAP3	345
001649	000		DO 840 I=1,NLOCT			QAP3	346
001650	000	840	WRITE(6,1)(D(I,J),J=1,NLOCT)			QAP3	347
001651	000		WRITE(6,27)ITER			QAP3	348
001652	000		WRITE(6,28)(IPASS(I),I=1,NFCIL)			QAP3	349
001653	000		WRITE(6,29)KEEP			QAP3	350
001654	000		WRITE(6,31)			QAP3	351
001655	000		WRITE(6,32)(LIXED(I),I=1,NFCIL)			QAP3	352
001656	000		DO 850 I=1,IEVAL			QAP3	353
001657	000	850	WRITE(6,33)I,ICOST(I),FOUND(I),(TLOCT(I,L),L=1,NFCIL)			QAP3	354
001658	000		WRITE(6,34)TIME			QAP3	355
001659	000	C				QAP3	356
001660	000		WRITE(6,07)			QAP3	357
001661	000		WRITE(6,42)			QAP3	358
001662	000		WRITE(6,43)KLBND			QAP3	359
001663	000		WRITE(6,44)KUPDT			QAP3	360
001664	000		WRITE(6,07)			QAP3	361
001665	000		WRITE(6,46)FTHRIO			QAP3	362
001666	000	C				QAP3	363
001667	000	C				QAP3	364
001668	000	C	CHECK WHETHER THERE ARE ANY MORE PROBLEMS TO SOLVE			QAP3	365
001669	000		900 IF(MORE)1000,1000,10			QAP3	366
001670	000	C				QAP3	367
001671	000		1000 STOP			QAP3	368
001672	000	C				QAP3	369
001673	000		1 FORMAT(10X,40I3)			QAP3	370
001674	000		2 FORMAT(10X,9('*** QAP 33 ***'))			QAP3	371
001675	000		3 FORMAT(20X,20I3)			QAP3	372
001676	000		4 FORMAT(1H1,' DATA FOR PROBLEM',I5/			QAP3	373
001677	000		* 10X,' NUMBER OF FACILITIES =',I3/			QAP3	374
001678	000		* 10X,' NUMBER OF LOCATIONS =',I3//)			QAP3	375

001679	000	5 FORMAT(10X,' DISTANCE FROM FACILITY')	QAP3 376
001680	000	6 FORMAT(1H1)	QAP3 377
001681	000	7 FORMAT(/)	QAP3 378
001682	000	8 FORMAT(/10X,'INTERACTION MATRIX'/)	QAP3 379
001683	000	9 FORMAT(20X,'AT ITERATION ',I5,' FACILITY',I3,' IS	QAP3 380
001684	000	*LOCATED',I3,' WITH AN LB COST IN COST LOGS',I5)	QAP3 381
001685	000	11 FORMAT(1H0,5X,'RESULTS OF THE MOST RECENT FEASIBLE SOLUTION , FOUND	QAP3 382
001686	000	*0 AT ITERATION',I5/)	QAP3 383
001687	000	12 FORMAT(20X,'LEVEL FACILITY LOCATION FIRST CMCST'/)	QAP3 384
001688	000	14 FORMAT(22X,I2,10X,I2,10X,I2,10X,I5)	QAP3 385
001689	000	16 FORMAT(10X,'END OF CALCULATION')	QAP3 386
001690	000	17 FORMAT(/10X,'STARTING LOWER BOUND ON THE TOTAL COST=',I5/)	QAP3 387
001691	000	21 FORMAT(18X,' ARRAY OF ALL THE DISTANCES FOLLOWING')	QAP3 388
001692	000	22 FORMAT(10X,' ARRAY OF ALL THE INTERFACILITIES FOLLOWING')	QAP3 389
001693	000	23 FORMAT(1H1,' SUMMARY OF THE ORIGINAL EX-PLAINED WITH PROBLEM'/)	QAP3 390
001694	000	24 FORMAT(10X,'NUMBER OF FACILITIES',I5)	QAP3 391
001695	000	26 FORMAT(10X,'NUMBER OF LB LOGS',I5)	QAP3 392
001696	000	27 FORMAT(/10X,'NUMBER OF THIS POINT ',I8)	QAP3 393
001697	000	28 FORMAT(/10X,'NUMBER OF POINTS AT VARIOUS LEVELS',I20)	QAP3 394
001698	000	29 FORMAT(/10X,'TOTAL NUMBER OF LB LOGS ',I5)	QAP3 395
001699	000	31 FORMAT(/50X,'LOCATION OF THE FACILITIES')	QAP3 396
001700	000	32 FORMAT(10X,'NUM , OF FACILITIES AT',I5,20X)	QAP3 397
001701	000	33 FORMAT(10X,I3,I7,I9,20X)	QAP3 398
001702	000	34 FORMAT(/10X,'TOTAL COST LOGS ',I10,2,' LOGS')	QAP3 399
001703	000	C	QAP3 400
001704	000	36 FORMAT(/10X,'AS A RESULT OF THE NEWLY FORMED FACILITY , WE HAVE ,	QAP3 401
001705	000	*LATEST LB=',I5,' COST LOGS ',I5,' LABEL LOGS',I5)	QAP3 402
001706	000	37 FORMAT(/10X,'HOWEVER , IT DOES NOT SEEM TO BE WORTH IT AS DIF=',	QAP3 403
001707	000	*1PE20.8)	QAP3 404
001708	000	38 FORMAT(/15X,'THE CURRENT LOWER BOUND UPPER BOUND ARE ',I2I7)	QAP3 405
001709	000	39 FORMAT(/10X,'SINCE THE CURRENT FACILITY WAS FORMED BETWEEN',I5,' + ',	QAP3 406
001710	000	*I5,' I WILL RESTART FROM THE LAST COMPLETED FACILITY')	QAP3 407
001711	000	41 FORMAT(02X,3(' FORCED TERMINATION *** '),I10,' AFTER ',I8,' ITERA	QAP3 408
001712	000	*TIONS')	QAP3 409
001713	000	C	QAP3 410
001714	000	42 FORMAT(10X,'NUMBER OF CALLS FOR SUBROUTINE',I5/)	QAP3 411
001715	000	43 FORMAT(18X,'SUBROUTINE ',I5,' WAS CALLED ',I5,' TIMES')	QAP3 412
001716	000	44 FORMAT(18X,'SUBROUTINE ',I5,' WAS CALLED ',I5,' TIMES')	QAP3 413
001717	000	46 FORMAT(10X,'FACILITY LABELS ',I5,'FB',I5,' PERCENT')	QAP3 414
001718	000	C	QAP3 415
001719	000	47 FORMAT(/10X,'NUMBER OF FACILITY ELEMENTS=',I5,' ARRANGED AS FOLLOWS	QAP3 416
001720	000	*/)	QAP3 417
001721	000	48 FORMAT(10X,'NUM',I5X,'ELEMENTS',I5X,'AT LOCATION'/)	QAP3 418
001722	000	49 FORMAT(10X,I2,9X,I2,10X,I5)	QAP3 419
001723	000	C	QAP3 420
001724	000	51 FORMAT(10X,'STARTING LOWER BOUND',I5/)	QAP3 421
001725	000	* 10X,'STARTING LOWER BOUND',I5/)	QAP3 422
001726	000	52 FORMAT(/10X,'NO. OF ITERATIONS SO FAR=',I5)	QAP3 423
001727	000	98 FORMAT(02X,4(' *** ERROR ***'),I10X,' IF I5.GT.1, IT EQUALS',	QAP3 424
001728	000	*I5)	QAP3 425
001729	000	99 FORMAT(2X,5(' ***NOTICE ***'),I5, IUBC=',I5,' ITOT=',I5)	QAP3 426
001730	000	C	QAP3 427
001731	000	END	QAP3 428
001732	000	DELTA,ST TUMER-S*SINAN-QAP5,UPDATE,,212330053012	
001733	000	SUBROUTINE UPDATE(I,UBC)	QAP3 505
001734	000	COMMON IC1,IC2,IC3,IPCL,IPCT,NP,M2,LOCST,ITOT	QAP3 506
001735	000	COMMON MSTER(04000),IMPRK(2000),KSNWS(60)	QAP3 507

001736	000	COMMON KLND,KUPDT	QAP3 508
001737	000	INTEGER DISTI(400),DISTJ(400),INTRI(400),INTRJ(400)	QAP3 509
001738	000	LOGICAL FCFLG(20),FLOGT(20),FLGDS(400),FLGIN(400)	QAP3 510
001739	000	C	QAP3 511
001740	000	EQUIVALENCE( MSTER(0801),DISTI(1))	QAP3 512
001741	000	EQUIVALENCE( MSTER(1201),DISTJ(1))	QAP3 513
001742	000	EQUIVALENCE( MSTER(1601),INTRI(1))	QAP3 514
001743	000	EQUIVALENCE( MSTER(2001),INTRJ(1))	QAP3 515
001744	000	C	QAP3 516
001745	000	EQUIVALENCE( MSTER(2501),FCFLG(1))	QAP3 517
001746	000	EQUIVALENCE( MSTER(2901),FLOGT(1))	QAP3 518
001747	000	EQUIVALENCE( MSTER(2601),FLGDS(1))	QAP3 519
001748	000	EQUIVALENCE( MSTER(3001),FLGIN(1))	QAP3 520
001749	000	C	QAP3 521
001750	000	C	QAP3 522
001751	000	C UPDATE THE NUMBER OF CALLS COUNTER	QAP3 523
001752	000	KUPDT=KUPDT+1	QAP3 524
001753	000	C	QAP3 525
001754	000	C	QAP3 526
001755	000	C **NOTATIONS	QAP3 527
001756	000	C I FACILITY	QAP3 528
001757	000	C J LOCATION	QAP3 529
001758	000	C KEY= 1 ...DISENGAGEMENT	QAP3 530
001759	000	C KEY= 2 ...ENGAGEMENT	QAP3 531
001760	000	C UNFLAG/FLAG THE ORDERED DISTANCES AND INTERACTIONS ACCORDING TO LAST	QAP3 532
001761	000	C DISENGAGEMENT/ASSIGNMENT	QAP3 533
001762	000	C A) LOCATION J HAS BEEN VACATED/ASSIGNED	QAP3 534
001763	000	DO 280 IK=1,N2	QAP3 535
001764	000	II=DISTI(IK)	QAP3 536
001765	000	JJ=DISTJ(IK)	QAP3 537
001766	000	GO TO(210,240),KEY	QAP3 538
001767	000	210 IF(J.EQ.II.AND..NOT.FLOGT(1))GO TO 220	QAP3 539
001768	000	IF(J.EQ.JJ.AND..NOT.FLOGT(1))GO TO 220	QAP3 540
001769	000	GO TO 280	QAP3 541
001770	000	220 FLGDS(IK)=.FALSE.	QAP3 542
001771	000	GO TO 280	QAP3 543
001772	000	240 IF(J.EQ.II.OR.J.EQ.JJ)GO TO 260	QAP3 544
001773	000	GO TO 280	QAP3 545
001774	000	260 FLGDS(IK)=.TRUE.	QAP3 546
001775	000	280 CONTINUE	QAP3 547
001776	000	C	QAP3 548
001777	000	C B) FACILITY I HAS BEEN DISENTANGLED/LOCATED	QAP3 549
001778	000	JK=J	QAP3 550
001779	000	DO 380 J=1,M2	QAP3 551
001780	000	II=INTRI(J)	QAP3 552
001781	000	JJ=INTRJ(J)	QAP3 553
001782	000	GO TO(310,340),KEY	QAP3 554
001783	000	310 IF(I.EQ.II.AND..NOT.FCFLG(JJ))GO TO 320	QAP3 555
001784	000	IF(I.EQ.JJ.AND..NOT.FCFLG(IJ))GO TO 320	QAP3 556
001785	000	GO TO 380	QAP3 557
001786	000	320 FLGIN(J)=.FALSE.	QAP3 558
001787	000	GO TO 380	QAP3 559
001788	000	340 IF(I.EQ.II. OR.I.EQ.JJ)GO TO 360	QAP3 560
001789	000	GO TO 380	QAP3 561
001790	000	360 FLGIN(J)=.TRUE.	QAP3 562
001791	000	380 CONTINUE	QAP3 563
001792	000	J=JK	QAP3 564

001793	000	C		QAP3 565
001794	000		RETURN	QAP3 566
001795	000		END	QAP3 567
001796	000		QELT,SI TUMER-S*SINAN-QAP5,LWBNB,,,212650053012	
001797	000	C	A SUBROUTINE TO CALCULATE THE LOWER BOUND AT ANY NODE	QAP3 568
001798	000		SUBROUTINE LWBND(LEVY,ITER,IWD,MOVE)	QAP3 569
001799	000		COMMON IC1,IC2,IC3,NFCIL,NLOCT,NP,M2,LRCST,ITOT	QAP3 570
001800	000		COMMON MSTER(04000),IWORK(2000),KSNSW(60)	QAP3 571
001801	000		COMMON KLBND,KUPDT	QAP3 572
001802	000		INTEGER F(20,20),D(20,20),LOCAT(20),LEVEL(20)	QAP3 573
001803	000		INTEGER WORK1(20,20),WORK2(20),WORK3(20),INDEX(20),JINDEX(20)	QAP3 574
001804	000		INTEGER DISTI(400),DISTJ(400),INTRI(400),INTRJ(400)	QAP3 575
001805	000		LOGICAL FCFLG(20),FLG5(400),BAND(20,20)	QAP3 576
001806	000		LOGICAL FLGDS(400),FLGID(400)	QAP3 577
001807	000	C		QAP3 578
001808	000		EQUIVALENCE( MSTER(0001),D(1,1))	QAP3 579
001809	000		EQUIVALENCE( MSTER(0002),D(2,1))	QAP3 580
001810	000		EQUIVALENCE( MSTER(0003),D(3,1))	QAP3 581
001811	000		EQUIVALENCE( MSTER(0004),D(4,1))	QAP3 582
001812	000		EQUIVALENCE( MSTER(0005),D(5,1))	QAP3 583
001813	000		EQUIVALENCE( MSTER(0006),D(6,1))	QAP3 584
001814	000	C		QAP3 585
001815	000		EQUIVALENCE( MSTER(0007),D(7,1))	QAP3 586
001816	000		EQUIVALENCE( MSTER(0008),D(8,1))	QAP3 587
001817	000		EQUIVALENCE( MSTER(0009),D(9,1))	QAP3 588
001818	000		EQUIVALENCE( MSTER(0010),D(10,1))	QAP3 589
001819	000	C		QAP3 590
001820	000		EQUIVALENCE( MSTER(0011),D(11,1))	QAP3 591
001821	000		EQUIVALENCE( MSTER(0012),D(12,1))	QAP3 592
001822	000		EQUIVALENCE( MSTER(0013),D(13,1))	QAP3 593
001823	000	C		QAP3 594
001824	000		EQUIVALENCE( IWORK(1000),D(14,1))	QAP3 595
001825	000	C		QAP3 596
001826	000	C	UPDATE THE NUMBER OF CALLS COUNTER	QAP3 597
001827	000		KLBND=KLBND+1	QAP3 598
001828	000	C		QAP3 599
001829	000		IF(KSNSW(21).EQ.1.AND.MOVE.EQ.0)WRITE(3,100)	QAP3 600
001830	000		IF(KSNSW(21).EQ.1.AND.MOVE.EQ.1)WRITE(3,100)LEVY	QAP3 601
001831	000		IF(KSNSW(21).EQ.1.AND.MOVE.EQ.2)WRITE(3,21)LEVY	QAP3 602
001832	000	C		QAP3 603
001833	000		IC1=IC1+IWD	QAP3 604
001834	000	C		QAP3 605
001835	000		IF(KSNSW(22).EQ.1)WRITE(6,22)IC1	QAP3 606
001836	000	C		QAP3 607
001837	000	C	CALCULATE THE IC2 MATRIX	QAP3 608
001838	000		260 IROW=0	QAP3 609
001839	000		IC2=0	QAP3 610
001840	000		IF(MOVE.NE.0)GO TO 270	QAP3 611
001841	000		IROW=NFCIL	QAP3 612
001842	000		GO TO 580	QAP3 613
001843	000		270 DO 400 I=1,NFCIL	QAP3 614
001844	000		IF(FCFLG(I))GO TO 400	QAP3 615
001845	000		IROW=IROW+1	QAP3 616
001846	000		WORK2(IROW)=32767	QAP3 617
001847	000		ICOL=0	QAP3 618
001848	000		DO 350 J=1,NLOCT	QAP3 619
001849	000		IF(FLAGT(J))GO TO 350	QAP3 620



001850	000	ICOL=ICOL+1	GAP3 621
001851	000	IF(.NOT.BANED(I,J))GO TO 280	GAP3 622
001852	000	KEEP=32767	GAP3 623
001853	000	GO TO 340	GAP3 624
001854	000	280 KEEP=0	GAP3 625
001855	000	DO 330 K=1,NFCIL	GAP3 626
001856	000	IF(.NOT.FCFLG(K))GO TO 330	GAP3 627
001857	000	L=LOCAT(K)	GAP3 628
001858	000	ISAV=F(I,K)	GAP3 629
001859	000	IF(F(K,I).GT.F(I,K))ISAV=F(K,I)	GAP3 630
001860	000	KEEP=KEEP+(ISAV*D(J,L))	GAP3 631
001861	000	330 CONTINUE	GAP3 632
001862	000	340 WORK1(IROW,ICOL)=KEEP	GAP3 633
001863	000	IF(KEEP.GE.WORK2(IROW))GO TO 350	GAP3 634
001864	000	WORK2(IROW)=KEEP	GAP3 635
001865	000	350 CONTINUE	GAP3 636
001866	000	IC2=IC2+WORK2(IROW)	GAP3 637
001867	000	400 CONTINUE	GAP3 638
001868	000	C	GAP3 639
001869	000	C WRITE THE IC2 MATRIC IF YOU WISH	GAP3 640
001870	000	IF(KSNSW(23).EQ.0)GO TO 460	GAP3 641
001871	000	KEEP=0	GAP3 642
001872	000	DO 430 I=1,NFCIL	GAP3 643
001873	000	IF(FCFLG(I))GO TO 430	GAP3 644
001874	000	KEEP=KEEP+1	GAP3 645
001875	000	INDEX(KEEP)=I	GAP3 646
001876	000	430 CONTINUE	GAP3 647
001877	000	KEEP=0	GAP3 648
001878	000	DO 450 J=1,NLOCT	GAP3 649
001879	000	IF(FLOGT(J))GO TO 450	GAP3 650
001880	000	KEEP=KEEP+1	GAP3 651
001881	000	JNDEX(KEEP)=J	GAP3 652
001882	000	450 CONTINUE	GAP3 653
001883	000	WRITE(6,23)	GAP3 654
001884	000	WRITE(6,11)(JNDEX(J),J=1,ICOL)	GAP3 655
001885	000	DO 460 I=1,IROW	GAP3 656
001886	000	460 WRITE(6,1)INDEX(I),(WORK1(I,J),J=1,ICOL),WORK2(I)	GAP3 657
001887	000	WRITE(6,7)	GAP3 658
001888	000	C CALCULATE THE MODIFIED MATRIX AND GET COLUMN MINIMA	GAP3 659
001889	000	480 DO 500 J=1,ICOL	GAP3 660
001890	000	WORK3(J)=32767	GAP3 661
001891	000	DO 490 I=1,IROW	GAP3 662
001892	000	WORK1(I,J)=WORK1(I,J)-WORK2(I)	GAP3 663
001893	000	IF(WORK1(I,J).GE.WORK3(J))GO TO 490	GAP3 664
001894	000	WORK3(J)=WORK1(I,J)	GAP3 665
001895	000	490 CONTINUE	GAP3 666
001896	000	500 CONTINUE	GAP3 667
001897	000	C	GAP3 668
001898	000	C PRINT COLUMN MINIMA	GAP3 669
001899	000	IF(KSNSW(23).EQ.0)GO TO 530	GAP3 670
001900	000	WRITE(6,11)(WORK3(J),J=1,ICOL)	GAP3 671
001901	000	C	GAP3 672
001902	000	C IN CASE ICOL.GT.IROW TAKE THE FIRST IROW COLUMNS AFTER RANKING	GAP3 673
001903	000	530 IF(ICOL.EQ.IROW)GO TO 540	GAP3 674
001904	000	CALL ARRNG(ICOL,1,1)	GAP3 675
001905	000	540 DO 550 J=1,IROW	GAP3 676
001906	000	550 IC2=IC2+WORK3(J)	GAP3 677

001907	000	C		QAP3 678
001908	000	C	WRITE RHE VALUES OF IC2 IF YOU WISH	QAP3 679
001909	000		IF(KSNSW(24).EQ.1)WRITE(6,24)IC2	QAP3 680
001910	000	C		QAP3 681
001911	000	C		QAP3 682
001912	000	C	CALCULATE THE VALUE OF IC3	QAP3 683
001913	000		580 IC3=0	QAP3 684
001914	000		IF(IROW.EQ.1)GO TO 700	QAP3 685
001915	000		IF(KSNSW(26).EQ.1)WRITE(6,26)	QAP3 686
001916	000		KEEP=0	QAP3 687
001917	000		K=1	QAP3 688
001918	000		IROW2=IROW*(IROW-1)/2	QAP3 689
001919	000		DO 650 I=1,N2	QAP3 690
001920	000		IF(FLGDS(I))GO TO 650	QAP3 691
001921	000		DO 620 J=K,M2	QAP3 692
001922	000		IF(I-LGIN(J))GO TO 620	QAP3 693
001923	000		KEEP=KEEP+1	QAP3 694
001924	000		II=UISTI(I)	QAP3 695
001925	000		IJ=UISTJ(I)	QAP3 696
001926	000		J1=INTRI(J)	QAP3 697
001927	000		JJ=INTRJ(J)	QAP3 698
001928	000		KAM=D(I1,IJ)*F(J1,JJ)	QAP3 699
001929	000		IC3=IC3+KAM	QAP3 700
001930	000		IF(KSNSW(26).EQ.1)WRITE(6,27)KEEP,I1,IJ,J1,JJ,D(I1,IJ),F(J1,JJ)	QAP3 701
001931	000		*KAM	QAP3 702
001932	000		GO TO 630	QAP3 703
001933	000		620 CONTINUE	QAP3 704
001934	000		630 IF(KEEP.E=.IROW2)GO TO 710	QAP3 705
001935	000		K=J+1	QAP3 706
001936	000		650 CONTINUE	QAP3 707
001937	000		DO 660 I=1,10	QAP3 708
001938	000		660 WRITE(6,98)KEEP,IROW,IROW2	QAP3 709
001939	000		STOP	QAP3 710
001940	000	C		QAP3 711
001941	000		700 IF(KSNSW(27).EQ.1)WRITE(6,28)IC3	QAP3 712
001942	000		IF(MOVE.NF.0)GO TO 720	QAP3 713
001943	000		KEEP=IC1+IC2+IC3	QAP3 714
001944	000		IF(KEEP.GT.LBCST)CONTINUE	QAP3 715
001945	000		WRITE(6,27)KEEP,LBCST	QAP3 716
001946	000		GO TO 750	QAP3 717
001947	000		720 LBCST=IC1+IC2+IC3	QAP3 718
001948	000		750 IF(KSNSW(28).EQ.1)WRITE(6,29)LBCST	QAP3 719
001949	000		IF(KSNSW(29).EQ.1)WRITE(6,30)IC1,IC2,IC3,LBCST	QAP3 720
001950	000	C		QAP3 721
001951	000	C		QAP3 722
001952	000		1 FORMAT(10X,20I5)	QAP3 723
001953	000		7 FORMAT(/)	QAP3 724
001954	000		11 FORMAT(15X,20I5/)	QAP3 725
001955	000		12 FORMAT(12X,I2,6X,I2,2X,I2,5X,I2,3X,I2,3(7X,I4))	QAP3 726
001956	000		14 FORMAT(90X,'VALUE OF IC3 =',I5/)	QAP3 727
001957	000		16 FORMAT(90X,'LOWER BOUND IC3 =',I5/)	QAP3 728
001958	000		17 FORMAT(/50X,'IC1=',I5,5X,'IC2=',I5,5X,'IC3=',I5,10X,'LBCST=',I7)	QAP3 729
001959	000		18 FORMAT(1H0,' RESULTS FOR CALCULATIONS FOR LB AT THE START')	QAP3 730
001960	000		19 FORMAT(1H0,' RESULTS FOR CALCULATIONS FOR LB AT ITER ',I3,'	QAP3 731
001961	000		*LEVEL',I3,' ON A FORWARD MOVE')	QAP3 732
001962	000			
001963	000			

001964	000		
001965	000	21 FORMAT(1H0,' RESULTS FOR CALCULATIONS FOR LB AT ITER ',I3,' QAP3 733	
001966	000	*LEVEL',I3,' ON A BACKWARD MOVE')	QAP3 734
001967	000	22 FORMAT(90X,'VALUE IF C1 =',I5,')	QAP3 735
001968	000	23 FORMAT(10X,'MATRIX FOR CALCULATING C2 FOLLOWS')	QAP3 736
001969	000	24 FORMAT(90X,'VALUE OF C2 =',I5,')	QAP3 737
001970	000	26 FORMAT(10X,'ARRAY OF ELEMENTS TO CALCULATE C3 FOLLOWS'//,	QAP3 738
001971	000	* 10X,'INDEX FROM/TO INTER BET. DIST. INTER VAL	QAP3 739
001972	000	*UE')	QAP3 740
001973	000	27 FORMAT(/10X,'CALCULATED STARTING LB=',I5,' SO WE TAKE LBCST=',	QAP3 741
001974	000	*I5)	QAP3 742
001975	000	98 FORMAT(02X,20(1H*),3(' ERROR IN LWBND \$\$',I5,' KEEP=',I3,' IROW=	QAP3 743
001976	000	*',I3,' IROW2=',I3)	QAP3 744
001977	000	RETURN	QAP3 745
001978	000	END	QAP3 746
001979	000	DELT,SI TUMER-S*SINAN-QAP5,MAX,,,213106053012	
001980	000	FUNCTION MAX(K,L)	QAP3 747
001981	000	MAX=K	QAP3 748
001982	000	IF(L.GT.K)MAX=L	QAP3 749
001983	000	RETURN	QAP3 750
001984	000	END	QAP3 751
001985	000	DELT,SI TUMER-S*SINAN-QAP5,ARRNG,,,195634053112	
001986	000	C A SUBROUTINE TO ARRANGE THE ARRAY	QAP3 429
001987	000	SUBROUTINE ARRNG(N,KEY,NNN)	QAP3 430
001988	000	COMMON IC1,IC2,IC3,NFCIL,NLOCT,M2,LBCST,ITOT	QAP3 431
001989	000	COMMON MSTER(04000),IWORK1(0400),KSNSW(60)	QAP3 432
001990	000	COMMON KLPID,KURDI	QAP3 433
001991	000	INTEGER F(20,20),L(20,20),DISTI(400),DISTJ(400),INTRI(400)	QAP3 434
001992	000	INTEGER INTRJ(400)	QAP3 435
001993	000	INTEGER VISTI(400),VISTJ(400),VALUE(400),WORKS(20)	QAP3 436
001994	000	EQUIVALENCE( MSTER(0001),F(1,1))	QAP3 437
001995	000	EQUIVALENCE( MSTER(0001),VISTI(1))	QAP3 438
001996	000	EQUIVALENCE( MSTER(0001),VISTJ(1))	QAP3 439
001997	000	EQUIVALENCE( MSTER(1001),F(1,1))	QAP3 440
001998	000	EQUIVALENCE( MSTER(1001),INTRI(1))	QAP3 441
001999	000	EQUIVALENCE( MSTER(2001),INTRJ(1))	QAP3 442
002000	000	C	QAP3 443
002001	000	EQUIVALENCE(IWORK1(0001),VISTI(001))	QAP3 444
002002	000	EQUIVALENCE(IWORK1(0401),VISTJ(001))	QAP3 445
002003	000	EQUIVALENCE(IWORK1(1201),VALUE(001))	QAP3 446
002004	000	EQUIVALENCE(IWORK1(1980),WORKS(001))	QAP3 447
002005	000	C	QAP3 448
002006	000	IF(KONY.EQ.1)KEEP=N	QAP3 449
002007	000	IF(KONY.EQ.1)GO TO 110	QAP3 450
002008	000	NNN=NFCIL	QAP3 451
002009	000	IF(KEY.EQ.1)NNN=NLOCT	QAP3 452
002010	000	KEEP=0	QAP3 453
002011	000	DO 100 I=1,NNN	QAP3 454
002012	000	DO 100 J=1,NNN	QAP3 455
002013	000	KEEP=KEEP+1	QAP3 456
002014	000	IF(KEY.EQ.1)VALUE(KEEP)=D(I,J)	QAP3 457
002015	000	IF(KEY.EQ.2)VALUE(KEEP)=F(I,J)	QAP3 458
002016	000	IF(KEY.EQ.1.AND.VALUE(KEEP).EQ.0)VALUE(KEEP)=32767	QAP3 459
002017	000	IF(KEY.EQ.1.AND.I.GT.J)VALUE(KEEP)=32767	QAP3 460
002018	000	VISTI(KEEP)=I	QAP3 461
002019	000	100 VISTJ(KEEP)=J	QAP3 462
002020	000	GO TO 120	QAP3 463

002021	000	110 DO 115 I=1,N	GAP3 464
002022	000	115 VALUE(I)=WORK3(I)	GAP3 465
002023	000	120 N1=N-1	GAP3 466
002024	000	DO 300 I=1,N1	GAP3 467
002025	000	J1=I+1	GAP3 468
002026	000	DO 200 J=J1,N	GAP3 469
002027	000	GO TO(130,150),KEY	GAP3 470
002028	000	130 IF(VALUE(I).LE.VALUE(J))GO TO 200	GAP3 471
002029	000	GO TO 160	GAP3 472
002030	000	150 IF(VALUE(I).GE.VALUE(J))GO TO 200	GAP3 473
002031	000	160 KEEP=VALUE(I)	GAP3 474
002032	000	VALUE(I)=VALUE(J)	GAP3 475
002033	000	VALUE(J)=KEEP	GAP3 476
002034	000	IF(KONY.EQ.1)GO TO 200	GAP3 477
002035	000	C	GAP3 478
002036	000	KEEP=VISTI(I)	GAP3 479
002037	000	VISTI(I)=VISTI(J)	GAP3 480
002038	000	VISTI(J)=KEEP	GAP3 481
002039	000	C	GAP3 482
002040	000	KEEP=VISTJ(I)	GAP3 483
002041	000	VISTJ(I)=VISTJ(J)	GAP3 484
002042	000	VISTJ(J)=KEEP	GAP3 485
002043	000	200 CONTINUE	GAP3 486
002044	000	300 CONTINUE	GAP3 487
002045	000	C	GAP3 488
002046	000	GO TO (300,310),KONY	GAP3 489
002047	000		
002048	000		
002049	000		
002050	000	GO TO(320,340),KEY	GAP3 491
002051	000	320 DISTI(I)=VISTI(I)	GAP3 492
002052	000	DISTJ(I)=VISTJ(I)	GAP3 493
002053	000	GO TO 350	GAP3 494
002054	000	340 INTRI(I)=VISTI(I)	GAP3 495
002055	000	INTRJ(I)=VISTJ(I)	GAP3 496
002056	000	350 CONTINUE	GAP3 497
002057	000	GO TO 500	GAP3 498
002058	000	C	GAP3 499
002059	000	380 DO 390 I=1,N	GAP3 500
002060	000	390 WORK3(I)=VALUE(I)	GAP3 501
002061	000	C	GAP3 502
002062	000	500 RETURN	GAP3 503
002063	000	END	GAP3 504
002064	000	QELT,SI TUMER-S*SINAN-QAP5.TIMOUT,,,163213053112	
002065	000	SUBROUTINE TIMOUT	
002066	000	COMMON IC1,IC2,IC3,DECT1,DECT,N2,M2,LPCST,ITOT	
002067	000	COMMON MSTER(04000),IMON(12000),PSNSW(60)	
002068	000	COMMON KLBND,KUPDT	
002069	000	K=ITIME(I,J)	
002070	000	T=1/5000.	
002071	000	WRITE(6,2000)T	
002072	000	2000 FORMAT(' TIME',F10.3,' SEC')	
002073	000	RETURN	
002074	000	END	
002075	000	QELT,SI TUMER-S*SINAN-QAP5.UTAIN,,,163704053112	
002076	000	C A SUBROUTINE TO READ DATA FOR THE QUADRATIC ASSIGNMENT PROBLEM	GAP3 752
002077	000	C THIS VERSION IS INTENDED FOR THE NUGENT ET AL DATA	GAP3 753

002076	000	SUBROUTINE DTAIN(JOB,IFRAC,NFIXD,ISTOP,MORE,ITERM)	GAP3 754
002079	000	COMMON IC1,IC2,IC3,NFCIL,NLGCT,N2,M2,LBCST,ITOT	GAP3 755
002080	000	COMMON MSTER(04000),IWORK(2000),KSNSW(60)	GAP3 756
002081	000	COMMON KLPND,KUPDT	GAP3 757
002082	000	INTEGER F(20,20),D(20,20),FIX(20),WHERE(20)	GAP3 758
002083	000	INTEGER WORK1(20,20)	
002084	000	EQUIVALENCE( MSTER(0001),F(1,1))	GAP3 760
002085	000	EQUIVALENCE( MSTER(0001),D(1,1))	GAP3 761
002086	000	EQUIVALENCE( MSTER(0001),FIX(1))	GAP3 762
002087	000	EQUIVALENCE( MSTER(0041),FIX(1))	GAP3 763
002088	000	EQUIVALENCE( IWORK(0001),WORK1(1,1))	GAP3 764
002089	000		GAP3 765
002090	000	C READ THE NUMBER OF FACILITIES , THE NUMBER OF LOCATIONS , THE FRACTI	GAP3 766
002091	000	C OF THE INTERVAL (ITOT-LNCT) , THE NUMBER OF FIXED FACILITIES , AND	GAP3 767
002092	000	C THE CONTINUATION PARAMETER	GAP3 768
002093	000	READ(5,1)JOB,NFCIL,NLGCT,IFRAC,NFIXD,MORE,ISTOP,ITERM,LBCST,ITOT	GAP3 769
002094	000	C	GAP3 770
002095	000	IF(ITOT.EQ.0)ITOT=3276700	GAP3 771
002096	000	DO 10 I=1,20	GAP3 772
002097	000	DO 10 J=1,20	GAP3 773
002098	000	F(I,J)=0	GAP3 774
002099	000	10 D(I,J)=0	GAP3 775
002100	000	C	GAP3 776
002101	000	DO 20 I=1,NFCIL	GAP3 777
002102	000	20 READ(5,2)(WORK1(I,J),J=1,NFCIL)	GAP3 778
002103	000	C	GAP3 779
002104	000	DO 30 I=1,NFCIL	GAP3 780
002105	000	DO 25 J=1,NFCIL	GAP3 781
002106	000	IF(I.EQ.J)GO TO 25	GAP3 782
002107	000	IF(I.GT.J)F(I,J)=WORK1(I,J)	GAP3 783
002108	000	IF(I.LT.J)D(I,J)=WORK1(I,J)	GAP3 784
002109	000	25 CONTINUE	GAP3 785
002110	000	30 CONTINUE	GAP3 786
002111	000	C	GAP3 787
002112	000	DO 40 I=1,NFCIL	GAP3 788
002113	000	DO 40 J=1,NFCIL	GAP3 789
002114	000	F(I,J)=F(J,I)	GAP3 790
002115	000	D(J,I)=D(I,J)	GAP3 791
002116	000	40 CONTINUE	GAP3 792
002117	000	C	GAP3 793
002118	000	DO 50 I=1,NFCIL	GAP3 794
002119	000	DO 45 J=1,NFCIL	GAP3 795
002120	000	IF(I.GT.J)F(I,J)=0	GAP3 796
002121	000	45 CONTINUE	GAP3 797
002122	000	50 CONTINUE	GAP3 798
002123	000	C	GAP3 799
002124	000	KSNSW(1)=1	GAP3 800
002125	000	KSNSW(4)=1	GAP3 801
002126	000	KSNSW(6)=1	GAP3 802
002127	000	KSNSW(07)=1	GAP3 803
002128	000	KSNSW(08)=1	GAP3 804
002129	000	C READ THE INDICES OF THE FACILITIES FIXED APRIORI	GAP3 805
002130	000	DO 70 I=1,NFIXD	GAP3 806
002131	000	70 READ(5,1) FIX(I),WHERE(I)	GAP3 807
002132	000	C	GAP3 808
002133	000	NFIXD=1	GAP3 809
002134	000	RETURN	GAP3 810

002135	000	C		QAP3 811
002136	000	C		QAP3 812
002137	000	C		QAP3 813
002138	000		1 FORMAT(20X,12I5)	QAP3 814
002139	000		2 FORMAT(10X,35I2)	QAP3 815
002140	000		END	QAP3 816
002141	000		DELT,SI TUMER-S* SINAN-QAP4,MATH,,200617060512	
002142	000		COMMON IC1,IC2,IC3,NFCIL,TEST,IP,M2,LRCST,ITOT	QAP4 1
002143	000		COMMON MSTER(04000),IWORK(0001),KSNSW(60)	QAP4 2
002144	000		COMMON KLKND,KUPDI	QAP4 3
002145	000		INTEGER F(20,20),D(20,20),DIST(400),DISTB(400),INTRI(400)	QAP4 4
002146	000		INTEGER INIRJ(400),I1V(400),LOCAT(20),FCLT(400),PTCST(20)	QAP4 5
002147	000		INTEGER FIXED(20),WOB(100),CNCST(20),L(100)	QAP4 6
002148	000		INTEGER IPASS(20),TCST(100),DOCT(100,20),DOCT(100)	QAP4 7
002149	000		INTEGER FIX(20),WOB(100)	QAP4 8
002150	000		LOGICAL FCFLG(20),FLOC(400),FI(400),FLOC(400),BANED(20,20)	QAP4 9
002151	000	C		QAP4 10
002152	000		LOGICAL WORK2(1300),WORK1(100)	QAP4 11
002153	000		INTEGER WORK3(0120),WORK4(0100),WORK6(0020)	QAP4 12
002154	000	C		QAP4 13
002155	000	C		QAP4 14
002156	000		EQUIVALENCE( MSTER(0001),WORK1(1))	QAP4 15
002157	000		EQUIVALENCE( MSTER(0002),WORK1(2))	QAP4 16
002158	000		EQUIVALENCE( MSTER(0003),WORK1(3))	QAP4 17
002159	000		EQUIVALENCE( MSTER(0004),WORK1(4))	QAP4 18
002160	000		EQUIVALENCE( MSTER(0005),WORK1(5))	QAP4 19
002161	000		EQUIVALENCE( MSTER(0006),WORK1(6))	QAP4 20
002162	000		EQUIVALENCE( MSTER(0007),WORK1(7))	QAP4 21
002163	000		EQUIVALENCE( MSTER(0008),WORK1(8))	QAP4 22
002164	000		EQUIVALENCE( MSTER(0009),WORK1(9))	QAP4 23
002165	000		EQUIVALENCE( MSTER(0010),WORK1(10))	QAP4 24
002166	000		EQUIVALENCE( MSTER(0011),WORK1(11))	QAP4 25
002167	000		EQUIVALENCE( MSTER(0012),WORK1(12))	QAP4 26
002168	000		EQUIVALENCE( MSTER(0013),WORK1(13))	QAP4 27
002169	000		EQUIVALENCE( MSTER(0014),WORK1(14))	QAP4 28
002170	000		EQUIVALENCE( MSTER(0015),WORK1(15))	QAP4 29
002171	000		EQUIVALENCE( MSTER(0016),WORK1(16))	QAP4 30
002172	000		EQUIVALENCE( MSTER(0017),WORK1(17))	QAP4 31
002173	000		EQUIVALENCE( MSTER(0018),WORK1(18))	QAP4 32
002174	000		EQUIVALENCE( MSTER(0019),WORK1(19))	QAP4 33
002175	000	C		QAP4 34
002176	000		EQUIVALENCE( IWORK(0001),WORK1(1,1))	QAP4 35
002177	000	C		QAP4 36
002178	000		EQUIVALENCE( MSTER(2001),WORK7(1))	QAP4 37
002179	000		EQUIVALENCE( MSTER(2002),WORK7(1))	QAP4 38
002180	000	C		QAP4 39
002181	000	C	READ THE KEYS SETTINGS FOR THE MAIN PROGRAM	QAP4 40
002182	000		READ(5,3)(KSNSW(1),I=01,40)	QAP4 41
002183	000	C	READ THE KEY SETTINGS FOR SUBROUTINE LWBND	QAP4 42
002184	000		READ(5,3)(KSNSW(1),I=01,20)	QAP4 43
002185	000	C		QAP4 44
002186	000	C	READ THE PROBLEM DATA	QAP4 45
002187	000		10 CALL DTAIN(JOB,IFKAC,HEIYD,ISOP,MORE,ITERM)	QAP4 46
002188	000	C		QAP4 47
002189	000	C	PRINT THE INITIAL DATA IF YOU WISH	QAP4 48
002190	000		IF(KSNSW(01).EQ.0)GO TO 110	QAP4 49
002191	000		WRITE(6,4)JOB,NFCIL,DOCT	QAP4 50

002192	000	DO 70 I=1,2	QAP4	51
002193	000	70 WRITE(6,2)	QAP4	52
002194	000	WRITE(6,7)	QAP4	53
002195	000	WRITE(6,5)	QAP4	54
002196	000	DO 80 I=1,NLOCT	QAP4	55
002197	000	80 WRITE(6,01)(D(I,J),J=1,NLOCT)	QAP4	56
002198	000	WRITE(6,8)	QAP4	57
002199	000	DO 90 I=1,NFCIL	QAP4	58
002200	000	90 WRITE(6,01)(F(I,J),J=1,NFCIL)	QAP4	59
002201	000	WRITE(6,7)	QAP4	60
002202	000	WRITE(6,51)LBCST,ITOT	QAP4	61
002203	000	C INITIALIZE THE COST	QAP4	62
002204	000	110 IC1=0	QAP4	63
002205	000	N2=NLOCT*NLOCT	QAP4	64
002206	000	M2=NFCIL*NFCIL	QAP4	65
002207	000	C UNFLAG ALL THE LOCATIONS AND THE FACILITIES	QAP4	66
002208	000	DO 130 I=1,NLOCT	QAP4	67
002209	000	130 FLOGT(I)=.FALSE.	QAP4	68
002210	000	DO 140 J=1,NFCIL	QAP4	69
002211	000	140 FCFLG(J)=.FALSE.	QAP4	70
002212	000	C	QAP4	71
002213	000	DO 150 I=1,NFCIL	QAP4	72
002214	000	IPASS(I)=0	QAP4	73
002215	000	DO 150 J=1,NLOCT	QAP4	74
002216	000	150 BAMED(I,J)=.FALSE.	QAP4	75
002217	000	C RANK THE DISTANCES IN AN ASCENDING ORDER	QAP4	76
002218	000	CALL ARRNG(N2,1,2)	QAP4	77
002219	000	C PRINT THE MATRIX OF ARRANGED DISTANCES	QAP4	78
002220	000	IF(KSNSW(02).EQ.0)GO TO 200	QAP4	79
002221	000	WRITE(6,21)	QAP4	80
002222	000	WRITE(6,01)(DISTI(I),I=1,N2)	QAP4	81
002223	000	WRITE(6,07)	QAP4	82
002224	000	WRITE(6,01)(DISTJ(I),I=1,M2)	QAP4	83
002225	000	C	QAP4	84
002226	000	C FLAG ALL THE DISTANCES	QAP4	85
002227	000	200 DO 210 I=1,N2	QAP4	86
002228	000	210 FLGDS(I)=.FALSE.	QAP4	87
002229	000	C	QAP4	88
002230	000	C RANK THE INTERACTIONS IN A DESCENDING ORDER	QAP4	89
002231	000	CALL ARRNG(M2,2,2)	QAP4	90
002232	000	C PRINT THE INTERACTIONS IF YOU WISH	QAP4	91
002233	000	IF(KSNSW(03).EQ.0)GO TO 310	QAP4	92
002234	000	WRITE(6,22)	QAP4	93
002235	000	WRITE(6,01)(INTRI(I),I=1,M2)	QAP4	94
002236	000	WRITE(6,07)	QAP4	95
002237	000	WRITE(6,01)(INTRJ(I),I=1,M2)	QAP4	96
002238	000	C	QAP4	97
002239	000	C FLAGG ALL THE INTERACTIONS	QAP4	98
002240	000	310 DO 320 I=1,M2	QAP4	99
002241	000	320 FLGIN(I)=.FALSE.	QAP4	100
002242	000	C INITIALIZE WORK7	QAP4	101
002243	000	DO 325 I=1,120	QAP4	102
002244	000	325 WORK7(I)=0	QAP4	103
002245	000	C INITIALIZE WORK8	QAP4	104
002246	000	DO 330 I=1,1300	QAP4	105
002247	000	330 WORK8(I)=.FALSE.	QAP4	106
002248	000	DO 335 I=1,20	QAP4	107

002249	000	355 CMCST(I)=0	GAP4 108
002250	000	C	GAP4 109
002251	000	C	GAP4 110
002252	000	TIME=0	GAP4 111
002253	000	LBTC=LBCST	GAP4 112
002254	000	IEVAL=0	GAP4 113
002255	000	IUBCS=0	GAP4 114
002256	000	KOMPL=0	GAP4 115
002257	000	FRAC=IFRAC	GAP4 116
002258	000	FRAC=0.01*FRAC	GAP4 117
002259	000	C	GAP4 118
002260	000	C INITIALIZE COUNTERS	GAP4 119
002261	000	KLBND=0	GAP4 120
002262	000	KUPDT=0	GAP4 121
002263	000	KFATH=0	GAP4 122
002264	000	KTERM=0	
002265	000	C	GAP4 123
002266	000	C	GAP4 124
002267	000	C	GAP4 125
002268	000	C A) MAKE THE FIXED ASSIGNMENTS FIRST	GAP4 126
002269	000	ITER=1	GAP4 127
002270	000	NFIXD=NFIXD	GAP4 128
002271	000	LEVY=1	GAP4 129
002272	000	PTCST(1)=0	GAP4 130
002273	000	IF(NFIXD.EQ.1)GO TO 425	GAP4 131
002274	000	DO 420 I=2,NFIXD	GAP4 132
002275	000	L2= FIX(I)	GAP4 133
002276	000	J=WHERE(I)	GAP4 134
002277	000	LEVEL(L2)=999	GAP4 135
002278	000	LOCAT(L2)=J	GAP4 136
002279	000	FCFLG(L2)=.TRUE.	GAP4 137
002280	000	FLOGT( J)=.TRUE.	GAP4 138
002281	000	DO 400 JJ=1,NLOCT	GAP4 139
002282	000	IF(JJ.EQ.J)GO TO 400	GAP4 140
002283	000	BANED(L2,JJ)=.TRUE.	GAP4 141
002284	000	400 CONTINUE	GAP4 142
002285	000	DO 405 JJ=1,NFCIL	GAP4 143
002286	000	IF(JJ.EQ.L2)GO TO 405	GAP4 144
002287	000	BANED(JJ,J)=.TRUE.	GAP4 145
002288	000	405 CONTINUE	GAP4 146
002289	000	CALL UPDATE(L2,J,2)	GAP4 147
002290	000	IF(KSNSW(05).EQ.1)WRITE(6,9)ITER,LEVY,L2,J	GAP4 148
002291	000	DO 410 JET=2,NFIXD	GAP4 149
002292	000	IF(JET.EQ.1)GO TO 410	GAP4 150
002293	000	JJ=WHERE(JET)	GAP4 151
002294	000	LL=FIX(JET)	GAP4 152
002295	000	ISAV=F(L2,LL)	GAP4 153
002296	000	IF(F(LL,L2).GT.ISAV)ISAV=F(LL,L2)	GAP4 154
002297	000	IC1=IC1+(ISAV*D(J,JJ))	GAP4 155
002298	000	410 CONTINUE	GAP4 156
002299	000	420 CONTINUE	GAP4 157
002300	000	IC1=0.5*IC1	GAP4 158
002301	000	IUBCS=IUBCS+IC1	GAP4 159
002302	000	C	GAP4 160
002303	000	C CALCULATE A STARTING LOWER BOUND	GAP4 161
002304	000	CALL LWBND(0,0,0,1)	GAP4 162
002305	000	GO TO 426	GAP4 163



002306	000	C		GAP4 164
002307	000		425 CALL LWBND(0,0,0,0)	GAP4 165
002308	000	C		GAP4 166
002309	000		426 IF(LBTC.LT.LBCST)LBTC=LBCLT	GAP4 167
002310	000		IF(KNSNW(04).EQ.1)WRITE(6,17)LBTC	GAP4 168
002311	000	C		GAP4 169
002312	000	C	B)PROCEED TO OBTAIN A FIRST FEASIBLE SOLUTION ALONG THE RIGHTMOST	GAP4 170
002313	000	C	BRANCH OR THE TREE	GAP4 171
002314	000		L2=FIX(1)	GAP4 172
002315	000		J=WHERE(1)	GAP4 173
002316	000		LEVEL(L2)=1	GAP4 174
002317	000		FIXED(1)=L2	GAP4 175
002318	000		LOCAT(L2)=J	GAP4 176
002319	000		LIXED(1)=J	GAP4 177
002320	000		JET=0	GAP4 178
002321	000		IF(NFIXD.EQ.1)GO TO 429	GAP4 179
002322	000		DO 427 I=2,NFIXD	GAP4 180
002323	000		JJ=WHERE(I)	GAP4 181
002324	000		LL= FIX(I)	GAP4 182
002325	000		ISAV=F(L2,LL)	GAP4 183
002326	000		IF(ISAV.LT.F(LL,L2))ISAV=F(LL,L2)	GAP4 184
002327	000		JET=JET+(ISAV*D(J,JJ))	GAP4 185
002328	000		427 CONTINUE	GAP4 186
002329	000		429 PTCST(1)=JET	GAP4 187
002330	000		CMCST(1)=JET+IC1	GAP4 188
002331	000		IUBCS=IUDCS+JET	GAP4 189
002332	000		FCLTY(1)=L2	GAP4 190
002333	000		FCFLG(L2)=.TRUE.	GAP4 191
002334	000		FLOGT(J)=.TRUE.	GAP4 192
002335	000		L1=J	GAP4 193
002336	000		CALL UPDATE(L2,L1,2)	GAP4 194
002337	000	C	B) CHOOSE THE ELEMENT WITH MAXIMUM INTERACTIONS WITH THE ONE ASSIGNED	GAP4 195
002338	000	C	AT THE PRECEDING LEVEL	GAP4 196
002339	000		430 LEVY=LEVY+1	GAP4 197
002340	000		IPASS(LEVY)=IPASS(LEVY)+1	GAP4 198
002341	000		ITER=ITER+1	GAP4 199
002342	000		JET=FCLTY(LEVY-1)	GAP4 200
002343	000		ISAV=-32767	GAP4 201
002344	000		DO 460 I=1,NFCIL	GAP4 202
002345	000		IF(FCFLG(I))GO TO 460	GAP4 203
002346	000		IF((F(I,JET).LE.ISAV.AND.F(JET,I).LE.ISAV)GO TO 460	GAP4 204
002347	000		ISAV=F(I,JET)	GAP4 205
002348	000		IF(F(JET,I).GT.F(I,JET))ISAV=F(JET,I)	GAP4 206
002349	000		KEEP=I	GAP4 207
002350	000		460 CONTINUE	GAP4 208
002351	000		470 FCFLG(KEEP)=.TRUE.	GAP4 209
002352	000		LEVEL(KEEP)=LEVY	GAP4 210
002353	000	C	C) ASSIGN THE ELEMENT TO A LOCATION SO THAT THE COST OF SUCH AN ASSIG	GAP4 211
002354	000	C	IS MINIMAL	GAP4 212
002355	000		510 IWD=32767	GAP4 213
002356	000		DO 530 J=1,NLOCT	GAP4 214
002357	000		IF(BANED(KEEP,J))GO TO 530	GAP4 215
002358	000		IF(FLOGT(J))GO TO 530	GAP4 216
002359	000		KAM=0	GAP4 217
002360	000		DO 520 I=1,NFCIL	GAP4 218
002361	000		IF(I.EQ.KEEP)GO TO 520	GAP4 219
002362	000		IF(.NOT.FCFLG(I))GO TO 520	GAP4 220

002363	000	L=LOCAT(I)	GAP4 221
002364	000	ISAVV=F(KEEP,I)	GAP4 222
002365	000	IF(F(I,KEEP).GT.F(KEEP,I))ISAVV=F(I,KEEP)	GAP4 223
002366	000	KAM=KAM*(ISAVV*D(J,L))	GAP4 224
002367	000	520 CONTINUE	GAP4 225
002368	000	IF(KAM.GT.IWD)GO TO 530	GAP4 226
002369	000	IWD=KAM	GAP4 227
002370	000	LOCAT(KEEP)=J	GAP4 228
002371	000	530 CONTINUE	GAP4 229
002372	000	IF(IWD.EQ.32767)GO TO 740	GAP4 230
002373	000	IF(KNSNW(05).EQ.1)WRITE(6,9)ITER,LEVY,KEEP,LOCAT(KEEP),IWD	GAP4 231
002374	000	J=LOCAT(KEEP)	GAP4 232
002375	000	FCLTY(LEVY)=KEEP	GAP4 233
002376	000	FLOGT(J)=.TRUE.	GAP4 234
002377	000	IUBCS=IUBCS+IWD	GAP4 235
002378	000	FIXED(LEVY)=KEEP	GAP4 236
002379	000	PTCST(LEVY)=IWD	GAP4 237
002380	000	CALL UPDATE(KEEP,J,2)	GAP4 238
002381	000	IF(KNSNW(06).EQ.0)GO TO 730	GAP4 239
002382	000	CMCST(LEVY)=IUBCS	GAP4 240
002383	000	LIXED(LEVY)=J	GAP4 241
002384	000	C	GAP4 242
002385	000	C CHECK WHETHER WE HAVE REACHED A TERMINAL NODE OR NOT	GAP4 243
002386	000	500 LEVX=LEVY+NFIXD-1	GAP4 244
002387	000	IF(LEVX.LT.NFCIL.AND.NFATH.EQ.0)GO TO 430	GAP4 245
002388	000	IF(LEVX.FO.NFCIL)GO TO 510	GAP4 246
002389	000	C CHECK WHETHER IT PAYS TO OPEN THE SEARCH ALONG THIS BRANCH	GAP4 247
002390	000	CALL LWBND(LEVY,ITER,IWD,1)	GAP4 248
002391	000	IF(LBCST.LT.ITOT)GO TO 575	GAP4 249
002392	000	KFATH=KFATH+1	GAP4 250
002393	000	GO TO 620	GAP4 251
002394	000	585 LEVY=LEVY+1	GAP4 252
002395	000	IPASS(LEVY)=IPASS(LEVY)+1	GAP4 253
002396	000	GO TO 730	GAP4 254
002397	000	C E) NOW WE HAVE A COMPLETE SOLUTION , PRINT IT IF YOU WISH	GAP4 255
002398	000	590 KOMPL=1	GAP4 256
002399	000	ICI=IUBCS	GAP4 257
002400	000	IF(IUBCS.GE.ITOT)GO TO 620	GAP4 258
002401	000	ITCT=IUBCS	GAP4 259
002402	000	IEVAL=IEVAL+1	GAP4 260
002403	000	ICOST(IEVAL)=IUBCS	GAP4 261
002404	000	FOUND(IEVAL)=ITER	GAP4 262
002405	000	DO 595 I=1,NFCIL	GAP4 263
002406	000	ILOCT(IEVAL,I)=LOCAT(I)	GAP4 264
002407	000	595 CONTINUE	GAP4 265
002408	000	600 IF(KNSNW(06).EQ.0)GO TO 610	GAP4 266
002409	000	WRITE(6,11)ITER	GAP4 267
002410	000	WRITE(6,12)	GAP4 268
002411	000	DO 605 I=1,LEVY	GAP4 269
002412	000	605 WRITE(6,14)I,FIXED(I),LIXED(I),PTCST(I),CMCST(I)	GAP4 270
002413	000	CALL TIMEOUT	
002414	000	C	GAP4 271
002415	000	C	GAP4 272
002416	000	C REDUCE THE CURRENT LOWER BOUND	GAP4 273
002417	000	610 DIF=ITOT-LBTC	GAP4 274
002418	000	KETOT=ITOT	GAP4 275
002419	000	KELBC=LBTC	GAP4 276

002420	000	IF(DIF.LE.ISTOP)KITER=100	
002421	000	IF(DIF.LE.ISTOP)GO TO 611	QAP4 277
002422	000	DIF=(FRAC*DIF)	QAP4 278
002423	000	ITOT=LBTC+DIF	QAP4 279
002424	000	611 IF(KSNSW(08).EQ.1)WRITE(6,35)KELC,ITOT,KETOT	
002425	000	C	QAP4 281
002426	000	C F) STORE THE CONFIGURATION OF THE CURRENT NODE SO THAT WE CAN RESTORE	QAP4 282
002427	000	C IT IF WE FAIL TO OBTAIN A BETTER SOLUTION UNDER THE CURRENT CEILING	QAP4 283
002428	000	DO 612 I=1,1300	QAP4 284
002429	000	WORK2(I)=.FALSE.	QAP4 285
002430	000	IF(WORK8(I))WORK2(I)=.TRUE.	QAP4 286
002431	000	612 CONTINUE	QAP4 287
002432	000	DO 615 I=1,120	QAP4 288
002433	000	615 WORK3(I)=WORK7(I)	QAP4 289
002434	000	DO 616 I=1,20	QAP4 290
002435	000	616 WORK6(I)=CMCST(I)	QAP4 291
002436	000	C	QAP4 292
002437	000	C *** BACKTRACK TO THE NEXT LEVEL UP. THE TREE	QAP4 293
002438	000	C A) FREE THE FACILITY AT THAT LEVEL	QAP4 294
002439	000	620 DO 640 I=1,NFCIL	QAP4 295
002440	000	IF(LEVEL(I).EQ.LEVY)GO TO 650	QAP4 296
002441	000	640 CONTINUE	QAP4 297
002442	000	C	QAP4 298
002443	000	650 FCFLG(I)=.FALSE.	QAP4 299
002444	000	DO 670 J=1,NLOCT	QAP4 300
002445	000	IF(J.EQ.LOCAT(I))GO TO 680	QAP4 301
002446	000	670 CONTINUE	QAP4 302
002447	000	680 FLOGT(J)=.FALSE.	QAP4 303
002448	000	IWD=-PTCST(LEVY)	QAP4 304
002449	000	IUBCS=IUBCS+IWD	QAP4 305
002450	000	BANED(I,J)=.TRUE.	QAP4 306
002451	000	C	QAP4 307
002452	000	C	QAP4 308
002453	000	C B) UPDATE THE FLAHS OF THE TWO ORDERED ARRAYS	QAP4 309
002454	000	CALL UPDATE(I,J,1)	QAP4 310
002455	000	C	QAP4 311
002456	000	C C) CALCULATE A LB ON ALL SOLUTIONS EXCLUDING THE LAST	QAP4 312
002457	000	CALL LWBND(LEVY,ITER,IWD,J)	QAP4 313
002458	000	C	QAP4 314
002459	000	C D) CHECK WHETHER IT PAYS TO RESUME THE SEARCH ALONG THIS BRANCH	QAP4 315
002460	000	IF(LBCST.LT.ITOT)GO TO 730	QAP4 316
002461	000	KFATH=KFATH+1	QAP4 317
002462	000	KEEP=I	QAP4 318
002463	000	GO TO 740	QAP4 319
002464	000	C	QAP4 320
002465	000	C RETRIEVE THE INDEX OF THE FACILITY AT THAT LEVEL	QAP4 321
002466	000	730 KEEP=FIXED(LEVY)	QAP4 322
002467	000	ITER=ITER+1	QAP4 323
002468	000	GO TO 470	QAP4 324
002469	000	C	QAP4 325
002470	000	740 DO 760 J=1,NLOCT	QAP4 326
002471	000	760 BANED(KEEP,J)=.FALSE.	QAP4 327
002472	000	FCFLG(KEEP)=.FALSE.	QAP4 328
002473	000	LEVY=LEVY-1	QAP4 329
002474	000	ITER=ITER+1	QAP4 330
002475	000	IF(LEVY.NE.0.AND.ITER.LT.ITERM)GO TO 620	QAP4 331
002476	000	IF(LEVY.EQ.0.AND.ITER.LT.ITERM)GO TO 775	QAP4 332

002477	000	WRITE(6,6)	GAP4 333
002478	000	WRITE(6,39)LBTC,ITOT	GAP4 334
002479	000	WRITE(6,7)	GAP4 335
002480	000	DO 765 I=1,2	GAP4 336
002481	000	765 WRITE(6,41)ITER	GAP4 337
002482	000	GO TO 810	GAP4 338
002483	000	C	GAP4 339
002484	000	C SINCE NO BETTER SOLUTION WAS FOUND , FLIP THE BOUNDS , RESTORE THE	GAP4 340
002485	000	C ARRAYS AND RETURN TO 620	GAP4 341
002486	000	775 IF(KSNSW(08).EQ.1)WRITE(6,39)LBTC,ITOT	GAP4 342
002487	000	IF(KSNSW(08).EQ.1)WRITE(6,42)ITER	GAP4 343
002488	000	CALL TIMEOUT	
002489	000	IF(KTERM.EQ.100)GO TO 800	
002490	000	LBTC=ITOT	GAP4 344
002491	000	KELBC=LBTC	GAP4 345
002492	000	IC1=KETOT	GAP4 346
002493	000	IUBCS=KETOT	GAP4 347
002494	000	ITOT=KETOT	GAP4 348
002495	000	DIF=ITOT-LBTC	GAP4 349
002496	000	IF(DIF.LT.ISTOP)GO TO 778	GAP4 350
002497	000	DIF=FRAC*DIF	GAP4 351
002498	000	GO TO 780	
002499	000	778 IF(KSNSW(08).EQ.1)WRITE(6,37)DIF	GAP4 352
002500	000	IF(KTERM.EQ.1)GO TO 800	
002501	000	KTERM=1	
002502	000	ITOT=KETOT	
002503	000	GO TO 782	
002504	000	C	GAP4 354
002505	000	780 KETOT=ITOT	GAP4 355
002506	000	ITOT=LBTC+DIF	GAP4 356
002507	000	782 DO 785 I=1,1300	GAP4 357
002508	000	WORK8(I)=.FALSE.	GAP4 358
002509	000	IF(WORK2(I))WORK8(I)=.TRUE.	GAP4 359
002510	000	785 CONTINUE	GAP4 360
002511	000	DO 790 I=1,120	GAP4 361
002512	000	790 WORK7(I)=WORK3(I)	GAP4 362
002513	000	DO 795 I=1,20	GAP4 363
002514	000	795 CMCST(I)=WORK6(I)	GAP4 364
002515	000	LEVY=NFCIL-NFIXD+1	GAP4 365
002516	000	C	GAP4 366
002517	000	IF(KSNSW(08).EQ.1)WRITE(6,38)LBTC,ITOT	GAP4 367
002518	000	GO TO 620	GAP4 368
002519	000	C	GAP4 369
002520	000	800 WRITE(6,16)	GAP4 370
002521	000	810 KEEP=0	GAP4 371
002522	000	FTHRIO=100.*FLOAT(KFATH)/FLOAT(KLBND)	GAP4 372
002523	000	DO 820 I=1,NFCIL	GAP4 373
002524	000	LIXED(I)=1	GAP4 374
002525	000	820 KEEP=KEEP+IPASS(I)	GAP4 375
002526	000	WRITE(6,23)	GAP4 376
002527	000	WRITE(6,24)NFCIL	GAP4 377
002528	000	WRITE(6,26)NLOCT	GAP4 378
002529	000	WRITE(6,07)	GAP4 379
002530	000	WRITE(6,47)NFIXD	GAP4 380
002531	000	WRITE(6,48)	GAP4 381
002532	000	DO 825 I=1,NFIXD	GAP4 382
002533	000	825 WRITE(6,49)I,FIX(I),WHERE(I)	GAP4 383

002534	000	WRITE(6,08)	QAP4 384
002535	000	DO 830 I=1,NFCIL	QAP4 385
002536	000	830 WRITE(6,1)(F(I,J),J=1,NFCIL)	QAP4 386
002537	000	WRITE(6,05)	QAP4 387
002538	000	DO 840 I=1,NLOCT	QAP4 388
002539	000	840 WRITE(6,1)(D(I,J),J=1,NLOCT)	QAP4 389
002540	000	WRITE(6,27)ITER	QAP4 390
002541	000	WRITE(6,28)(IPASS(I),I=1,NFCIL)	QAP4 391
002542	000	WRITE(6,29)KEEP	QAP4 392
002543	000	WRITE(6,31)	QAP4 393
002544	000	WRITE(6,32)(LIXED(I),I=1,NFCIL)	QAP4 394
002545	000	DO 850 I=1,IEVAL	QAP4 395
002546	000	850 WRITE(6,33)I,ICOST(I),FOUND(I),(TLOCT(I,L),L=1,NFCIL)	QAP4 396
002547	000	WRITE(6,34)TIME	QAP4 397
002548	000	C	QAP4 398
002549	000	WRITE(6,07)	QAP4 399
002550	000	WRITE(6,42)	QAP4 400
002551	000	WRITE(6,43)KLBND	QAP4 401
002552	000	WRITE(6,44)KUPDT	QAP4 402
002553	000	WRITE(6,07)	QAP4 403
002554	000	WRITE(6,46)FTHRIO	QAP4 404
002555	000	C	QAP4 405
002556	000	C	QAP4 406
002557	000	C CHECK WHETHER THERE ARE ANY MORE PROBLEMS TO SOLVE	QAP4 407
002558	000	IF(MORE)1000,1000,10	QAP4 408
002559	000	C	QAP4 409
002560	000	1000 STOP	QAP4 410
002561	000	C	QAP4 411
002562	000	1 FORMAT(10X,40I3)	QAP4 412
002563	000	2 FORMAT(10X,9('*** QAP 4 -'))	QAP4 413
002564	000	3 FORMAT(20X,20I3)	QAP4 414
002565	000	4 FORMAT(1H1,' DATA FOR PROBLEM',I5/	QAP4 415
002566	000	* 10X,' NUMBER OF FACILITIES =',I3/	QAP4 416
002567	000	* 10X,' NUMBER OF LOCATIONS =',I3//)	QAP4 417
002568	000	5 FORMAT(10X,' DISTANCE MATRIX',I/)	QAP4 418
002569	000	6 FORMAT(1H1)	QAP4 419
002570	000	7 FORMAT(/)	QAP4 420
002571	000	8 FORMAT(/10X,'INTERACTION MATRIX',/)	QAP4 421
002572	000	9 FORMAT(20X,'AT ITERATION',I,' LEVEL',I3,' FACILITY',I3,' IS	QAP4 422
002573	000	*LOCATED',I3,' WITH AN INCREASE IN COST EQUALS',I5)	QAP4 423
002574	000	11 FORMAT(1H0,5X,'RESULTS OF THE MOST RECENT FEASIBLE SOLUTION , FOUND	QAP4 424
002575	000	*D AT ITERATION',I6/)	QAP4 425
002576	000	12 FORMAT(20X,'LEVEL FACILITY LOCATION FCOST CMCST',/)	QAP4 426
002577	000	14 FORMAT(22X,I2,10X,I2,9X,I3,5X,I4,4X,I6)	QAP4 427
002578	000	16 FORMAT(10X,'END OF CALCULATIONS')	QAP4 428
002579	000	17 FORMAT(/10X,'STARTING LOWER BOUND ON THE TOTAL COST=',I5/)	QAP4 429
002580	000	21 FORMAT(10X,' ARRAY OF ASSIGNED DISTANCES FOLLOWS',/)	QAP4 430
002581	000	22 FORMAT(10X,' ARRAY OF ASSIGNED INTERACTIONS FOLLOWS')	QAP4 431
002582	000	23 FORMAT(1H1,' SUMMARY OF COMPUTATIONAL EXPERIENCE WITH PROBLEM',/)	QAP4 432
002583	000	24 FORMAT(10X,'NUMBER OF FACILITIES =',I5)	QAP4 433
002584	000	26 FORMAT(10X,'NUMBER OF LOCATIONS =',I5)	QAP4 434
002585	000	27 FORMAT(/10X,'NUMBER OF ITERATIONS =',I8)	QAP4 435
002586	000	28 FORMAT(/10X,'NUMBER OF NODES AT VARIOUS LEVELS',20I4)	QAP4 436
002587	000	29 FORMAT(/10X,'TOTAL NUMBER OF NODES =',I8)	QAP4 437
002588	000	31 FORMAT(/50X,'LOCATIONS OF THE FACILITIES')	QAP4 438
002589	000	32 FORMAT( 10X,'NUM , COST',I FOUND AT',2X,20I3/)	QAP4 439
002590	000	33 FORMAT(10X,I3,I7,I9,3X,20I3)	QAP4 440

002591	000	34 FORMAT(/10X,'TOTAL CPU TIME =',F10.2,' SECONDS')	GAP4 441
002592	000		GAP4 442
002593	000	C 36 FORMAT(/10X,'AS A RESULT OF THIS NEWLY FOUND SOLUTION , WE HAVE ,	GAP4 443
002594	000	*LATEST LB=,I5,' (CURRENT LB=,I5,' (LAST COMB=,I5)	GAP4 444
002595	000	37 FORMAT(/10X,'HOWEVER , IT DID NOT SEEM TO BE WORTH IT AS DIF=,	GAP4 445
002596	000	*1PE20.8)	GAP4 446
002597	000	38 FORMAT(/15X,'THE CURRENT LOWER AND UPPER BOUNDS ARE ',2I7)	GAP4 447
002598	000	39 FORMAT(/10X,'SINCE NO BETTER SOLUTION WAS FOUND BETWEEN',I5,' + ',	GAP4 448
002599	000	*I5,' I WILL RESTART FROM THE LAST COMBINATION FOUND')	GAP4 449
002600	000	41 FORMAT(2X,3(' FORMED COMBINATION *** ',I5,' AFTER ',I5,' ITERA-	GAP4 450
002601	000	*TIONS')	GAP4 451
002602	000	C 42 FORMAT(10X,'NUMBER OF CALLS TO SUBROUTINE 107)	GAP4 452
002603	000	43 FORMAT(10X,'SUBROUTINE 107 WAS CALLED ',I5,' TIMES')	GAP4 453
002604	000	44 FORMAT(10X,'SUBROUTINE 107 WAS CALLED ',I5,' TIMES')	GAP4 454
002605	000	46 FORMAT(10X,'FATHOMING TIME FRACTION=,F8.1,' PER CENT')	GAP4 455
002606	000	C 47 FORMAT(/10X,'NUMBER OF LEAD ELEMENTS=,I5,' ARRANGED AS FOLLOWS,	GAP4 456
002607	000	*/)	GAP4 457
002608	000	48 FORMAT(10X,'NUM',I5,' LEAD ELEMENTS AT LOCATION',I7)	GAP4 458
002609	000	49 FORMAT(10X,I2,9X,I2,I10,1X)	GAP4 459
002610	000	C 51 FORMAT(10X,'STARTING LOWER BOUND=,I8/	GAP4 460
002611	000	* 10X,'STARTING UPPER BOUND=,I8/)	GAP4 461
002612	000	52 FORMAT(/10X,'NO. OF ITERATIONS NO FAIL=,I5)	GAP4 462
002613	000	C 99 FORMAT(2X,5(' ***ITERATION ***'),' ITERATION=,I5,' ITOT=,I5)	GAP4 463
002614	000		GAP4 464
002615	000	C	GAP4 465
002616	000	END	GAP4 466
002617	000	DELTA,SI TUMER-S+SIHAN-GAP4 467	GAP4 467
002618	000		GAP4 468
002619	000		GAP4 469
002620	000	C A SUBROUTINE TO ARRANGE THE LEAD	GAP4 470
002621	000	SUBROUTINE ARRANGING THE LEAD	GAP4 471
002622	000	COMMON IC1,IC2,IC3,IC4,IC5,IC6,IC7,IC8,IC9,IC10,IC11,IC12,IC13,IC14,IC15	GAP4 472
002623	000	COMMON MSTER(10000),MVAL(1000),KNSM(100)	GAP4 473
002624	000	COMMON KLND,KLNDT	GAP4 474
002625	000	INTEGER F(20,20),D(20,20),V(400),DIS(1000),INTRI(400)	GAP4 475
002626	000	INTEGER INTRJ(400)	GAP4 476
002627	000	INTEGER VISTI(400),VISTJ(400),VALUE(400),KND(20)	GAP4 477
002628	000	EQUIVALENCE( MSTER(10000),F(1,1))	GAP4 478
002629	000	EQUIVALENCE( MSTER(10000),D(1,1))	GAP4 479
002630	000	EQUIVALENCE( MSTER(10000),V(1,1))	GAP4 480
002631	000	EQUIVALENCE( MSTER(10000),DIS(1,1))	GAP4 481
002632	000	EQUIVALENCE( MSTER(10000),INTRI(1,1))	GAP4 482
002633	000	EQUIVALENCE( MSTER(10000),INTRJ(1,1))	GAP4 483
002634	000	C EQUIVALENCE( MSTER(10000),KND(1,1))	GAP4 484
002635	000	EQUIVALENCE( IWORK1(1000),VISTI(101))	GAP4 485
002636	000	EQUIVALENCE( IWORK1(1001),VISTJ(101))	GAP4 486
002637	000	EQUIVALENCE( IWORK1(1002),VALUE(101))	GAP4 487
002638	000	EQUIVALENCE( IWORK1(1003),KND(1,1))	GAP4 488
002639	000	C	GAP4 489
002640	000	IF(KONY.EQ.1)KEEP=N	GAP4 490
002641	000	IF(KONY.EQ.1)GO TO 110	GAP4 491
002642	000	NNN=NFCIL	GAP4 492
002643	000	IF(KEY.EQ.1)NNN=NLOCT	GAP4 493
002644	000	KEEP=0	GAP4 494
002645	000	DO 100 I=1,NNN	GAP4 495
002646	000	DO 100 J=1,NNN	GAP4 496
002647	000		

002648	000	KEEP=KEEP+1	GAP4 497
002649	000	IF(KEY.EQ.1)VALUE(KEEP)=DIST(I,J)	GAP4 498
002650	000	IF(KEY.EQ.2)VALUE(KEEP)=DIST(I,J)	GAP4 499
002651	000	IF(KEY.EQ.1.AND.VALUE(KEEP).EQ.0)VALUE(KEEP)=32767	GAP4 500
002652	000	IF(KEY.EQ.1.AND.I.GT.J)VALUE(KEEP)=32767	GAP4 501
002653	000	VISTI(KEEP)=I	GAP4 502
002654	000	100 VISTJ(KEEP)=J	GAP4 503
002655	000	GO TO 120	GAP4 504
002656	000	110 DO 115 I=1,N	GAP4 505
002657	000	115 VALUE(I)=WORK3(I)	GAP4 506
002658	000	120 N1=N-1	GAP4 507
002659	000	DO 300 I=1,N1	GAP4 508
002660	000	J1=I+1	GAP4 509
002661	000	DO 200 J=J1,N	GAP4 510
002662	000	GO TO(130,150),KEY	GAP4 511
002663	000	130 IF(VALUE(I).LE.VALUE(J))GO TO 200	GAP4 512
002664	000	GO TO 160	GAP4 513
002665	000	150 IF(VALUE(I).GE.VALUE(J))GO TO 200	GAP4 514
002666	000	160 KEEP=VALUE(I)	GAP4 515
002667	000	VALUE(I)=VALUE(J)	GAP4 516
002668	000	VALUE(J)=KEEP	GAP4 517
002669	000	IF(KONY.EQ.1)GO TO 200	GAP4 518
002670	000	C	GAP4 519
002671	000	KEEP=VISTI(I)	GAP4 520
002672	000	VISTI(I)=VISTI(J)	GAP4 521
002673	000	VISTI(J)=KEEP	GAP4 522
002674	000	C	GAP4 523
002675	000	KEEP=VISTJ(I)	GAP4 524
002676	000	VISTJ(I)=VISTJ(J)	GAP4 525
002677	000	VISTJ(J)=KEEP	GAP4 526
002678	000	200 CONTINUE	GAP4 527
002679	000	300 CONTINUE	GAP4 528
002680	000	C	GAP4 529
002681	000	GO TO (380,310),KONY	GAP4 530
002682	000	310 DO 350 I=1,N	GAP4 531
002683	000	GO TO(320,340),KEY	GAP4 532
002684	000	320 DISTI(I)=VISTI(I)	GAP4 533
002685	000	DISTJ(I)=VISTJ(I)	GAP4 534
002686	000	GO TO 350	GAP4 535
002687	000	340 INTRI(I)=VISTI(I)	GAP4 536
002688	000	INTRJ(I)=VISTJ(I)	GAP4 537
002689	000	350 CONTINUE	GAP4 538
002690	000	GO TO 500	GAP4 539
002691	000	C	GAP4 540
002692	000	380 DO 390 I=1,N	GAP4 541
002693	000	390 WORK3(I)=VALUE(I)	GAP4 542
002694	000	C	GAP4 543
002695	000	500 RETURN	GAP4 544
002696	000	END	GAP4 545
002697	000	DELT,SI TUMER-S*SINAN-GAP4.UPDATE,,,234317053012	
002698	000	SUBROUTINE UPDATE(I,J,KEY)	GAP4 546
002699	000	COMMON IC1,IC2,IC3,HCFLG,LOC1,M2,M2,LRCST,ITOT	GAP4 547
002700	000	COMMON MSTP(04000),IWORK(2000),XSNSW(60)	GAP4 548
002701	000	COMMON KLAND,KUPDT	GAP4 549
002702	000	INTEGER DISTI(400),DISTJ(400),INTRI(400),INTRJ(400)	GAP4 550
002703	000	LOGICAL FCFLG(20),FLOGT(20),FLGDS(400),FLGIN(400)	GAP4 551
002704	000	C	GAP4 552

002705	000	EQUIVALENCE( MSTER(0001),DISTI(1))	QAP4 553
002706	000	EQUIVALENCE( MSTER(0001),DISTJ(1))	QAP4 554
002707	000	EQUIVALENCE( MSTER(0001),INTRI(1))	QAP4 555
002708	000	EQUIVALENCE( MSTER(0001),INTRJ(1))	QAP4 556
002709	000	C	QAP4 557
002710	000	EQUIVALENCE( MSTER(0001),FLGDS(1))	QAP4 558
002711	000	EQUIVALENCE( MSTER(0001),FLGDS(1))	QAP4 559
002712	000	EQUIVALENCE( MSTER(0001),FLGDS(1))	QAP4 560
002713	000	EQUIVALENCE( MSTER(0001),FLGDS(1))	QAP4 561
002714	000	C	QAP4 562
002715	000	C UPDATE THE COUNTER FOR THE NUMBER OF CALLS	QAP4 563
002716	000	KUPDT=KUPDT+1	QAP4 564
002717	000	C	QAP4 565
002718	000	C	QAP4 566
002719	000	C **NOTATIONS	QAP4 567
002720	000	C I FACILITY	QAP4 568
002721	000	C J LOCATION	QAP4 569
002722	000	C KEY= 1 ...DISENGAGEMENT	QAP4 570
002723	000	C KEY= 2 ...ENGAGEMENT	QAP4 571
002724	000	C UNFLAG/FLAG THE ORIGIN DISTANCES AND INTERACTIONS ACCORDING TO LAST	QAP4 572
002725	000	C DISENGAGEMENT/ASSIGNMENT	QAP4 573
002726	000	C A) LOCATION J HAS BEEN VARIATED/ASSIGNED	QAP4 574
002727	000	DO 280 IK=1,M2	QAP4 575
002728	000	II=DISTI(IK)	QAP4 576
002729	000	JJ=DIJ(IK)	QAP4 577
002730	000	GO TO(210,240),KEY	QAP4 578
002731	000	210 IF(J.EQ.II.AND..NOT.FLGDS(JJ))GO TO 220	QAP4 579
002732	000	IF(J.EQ.II.AND..NOT.FLGDS(JJ))GO TO 220	QAP4 580
002733	000	GO TO 280	QAP4 581
002734	000	220 FLGDS(IK)=.FALSE.	QAP4 582
002735	000	GO TO 280	QAP4 583
002736	000	240 IF(J.EQ.II.OR.J.EQ.II)GO TO 260	QAP4 584
002737	000	GO TO 280	QAP4 585
002738	000	260 FLGDS(IK)=.TRUE.	QAP4 586
002739	000	260 CONTINUE	QAP4 587
002740	000	C	QAP4 588
002741	000	C B) FACILITY I HAS BEEN DISTANGLED/LOCATED	QAP4 589
002742	000	JK=J	QAP4 590
002743	000	DO 380 J=1,M2	QAP4 591
002744	000	II=INTRI(J)	QAP4 592
002745	000	JJ=INTRJ(J)	QAP4 593
002746	000	GO TO(310,340),KEY	QAP4 594
002747	000	310 IF(I.EQ.II.AND..NOT.FLGDS(JJ))GO TO 320	QAP4 595
002748	000	IF(I.EQ.II.AND..NOT.FLGDS(JJ))GO TO 320	QAP4 596
002749	000	GO TO 380	QAP4 597
002750	000	320 FLGIN(J)=.FALSE.	QAP4 598
002751	000	GO TO 380	QAP4 599
002752	000	340 IF(I.EQ.II. OR.I.EQ.II)GO TO 360	QAP4 600
002753	000	GO TO 380	QAP4 601
002754	000	360 FLGIN(J)=.TRUE.	QAP4 602
002755	000	360 CONTINUE	QAP4 603
002756	000	J=JK	QAP4 604
002757	000	C	QAP4 605
002758	000	RETURN	QAP4 606
002759	000	END	QAP4 607
002760	000	DELTA,SI TUMER-S*SINAN-QAP4.LWIND,,,235321053012	
002761	000	C A SUBROUTINE TO CALCULATE THE LOWER BOUND AT ANY NODE	QAP4 608



002762	000	SUBROUTINE LWBND(LEVY,ITER,MOVE)	QAP4 609
002763	000	COMMON IC1,IC2,IC3,NFCIL,N2,M2,LRCST,ITOT	QAP4 610
002764	000	COMMON MSTER(04000),IWORK(32000),KSNSW(60)	QAP4 611
002765	000	COMMON KLPND,KUPDT	QAP4 612
002766	000	INTEGER F(20,20),D(20,20),MST(20),LEVFL(20)	QAP4 613
002767	000	INTEGER WORK1(20,20),WORK3(20),ITOT(20),JINDEX(20)	QAP4 614
002768	000	INTEGER DIST1(400),DIST2(400),IRI(400),ISIRI(400)	QAP4 615
002769	000	LOGICAL FCFLG(20),FLOGT(20),BANED(20,20)	QAP4 616
002770	000	LOGICAL FLGDS(400),FLGDI(400)	QAP4 617
002771	000	C	QAP4 618
002772	000	EQUIVALENCE( MSTER(000),F(1,1))	QAP4 619
002773	000	EQUIVALENCE( MSTER(000),D(1,1))	QAP4 620
002774	000	EQUIVALENCE( MSTER(000),MST(1))	QAP4 621
002775	000	EQUIVALENCE( MSTER(100),LEVFL(1))	QAP4 622
002776	000	EQUIVALENCE( MSTER(100),DIST1(1))	QAP4 623
002777	000	EQUIVALENCE( MSTER(200),DIST2(1))	QAP4 624
002778	000	C	QAP4 625
002779	000	EQUIVALENCE( MSTER(300),FLOGT(1))	QAP4 626
002780	000	EQUIVALENCE( MSTER(300),BANED(1))	QAP4 627
002781	000	EQUIVALENCE( MSTER(200),LEVFL(1))	QAP4 628
002782	000	EQUIVALENCE( MSTER(200),FLOGT(1))	QAP4 629
002783	000	C	QAP4 630
002784	000	EQUIVALENCE( MSTER(300),LEVFL(1))	QAP4 631
002785	000	EQUIVALENCE( MSTER(300),BANED(1))	QAP4 632
002786	000	EQUIVALENCE( MSTER(300),BANED(1,1))	QAP4 633
002787	000	C	QAP4 634
002788	000	EQUIVALENCE( IWORK(1000),JINDEX(001))	QAP4 635
002789	000	C	QAP4 636
002790	000	C	QAP4 637
002791	000	C UPDATE THE COUNTER FOR THE NUMBER OF CALLS	QAP4 638
002792	000	KLBNB=KLBNB+1	QAP4 639
002793	000	C	QAP4 640
002794	000	IF(KSNSW(21).EQ.1.AND.MOVE(10,0))WRITE(6,10)	QAP4 641
002795	000	IF(KSNSW(21).EQ.1.AND.MOVE(10,1))WRITE(6,10)ITER,LEVY	QAP4 642
002796	000	IF(KSNSW(21).EQ.1.AND.MOVE(10,2))WRITE(6,21)ITER,LEVY	QAP4 643
002797	000	C	QAP4 644
002798	000	IC1=IC1+IWD	QAP4 645
002799	000	C	QAP4 646
002800	000	IF(KSNSW(22).EQ.1)WRITE(6,22)IC1	QAP4 647
002801	000	C	QAP4 648
002802	000	C CALCULATE THE IC2 MATRIX	QAP4 649
002803	000	200 IROW=0	QAP4 650
002804	000	IC2=0	QAP4 651
002805	000	IF(MOVE.NE.0)GO TO 270	QAP4 652
002806	000	IROW=NFCIL	QAP4 653
002807	000	GO TO 580	QAP4 654
002808	000	270 DO 400 I=1,NFCIL	QAP4 655
002809	000	IF(FCFLG(I))GO TO 400	QAP4 656
002810	000	IROW=IROW+1	QAP4 657
002811	000	WORK2(IROW)=32767	QAP4 658
002812	000	ICOL=0	QAP4 659
002813	000	DO 350 J=1,NLOCT	QAP4 660
002814	000	IF(FLOGT(J))GO TO 350	QAP4 661
002815	000	ICOL=ICOL+1	QAP4 662
002816	000	IF(.NOT.BANED(I,J))GO TO 280	QAP4 663
002817	000	KEEP=32767	QAP4 664
002818	000	GO TO 340	QAP4 665

002819	000	280	KEEP=0	QAP4	666
002820	000		DO 330 K=1,NFCIL	QAP4	667
002821	000		IF(.NOT.FCFLG(K))GO TO 330	QAP4	668
002822	000		L=LOCAT(K)	QAP4	669
002823	000		ISAV=F(I,K)	QAP4	670
002824	000		IF(F(K,I).GT.F(I,K))ISAV=F(K,I)	QAP4	671
002825	000		KEEP=KEEP+(ISAV=0(J,I))	QAP4	672
002826	000	330	CONTINUE	QAP4	673
002827	000	340	WORK1(IROW,ICOL)=KEEP	QAP4	674
002828	000		IF(KEEP.GF.WORK2(IROW))GO TO 350	QAP4	675
002829	000		WORK2(IROW)=KEEP	QAP4	676
002830	000	350	CONTINUE	QAP4	677
002831	000		IC2=IC2+WORK2(IROW)	QAP4	678
002832	000	400	CONTINUE	QAP4	679
002833	000	C		QAP4	680
002834	000	C	WRITE THE IC2 MATRIX IF YOU WISH	QAP4	681
002835	000		IF(KSNSW(23).EQ.0)GO TO 430	QAP4	682
002836	000		KEEP=0	QAP4	683
002837	000		DO 430 I=1,NFCIL	QAP4	684
002838	000		IF(FCFLG(I))GO TO 430	QAP4	685
002839	000		KEEP=KEEP+1	QAP4	686
002840	000		INDEX(KEEP)=I	QAP4	687
002841	000	430	CONTINUE	QAP4	688
002842	000		KEEP=0	QAP4	689
002843	000		DO 450 J=1,NLOCT	QAP4	690
002844	000		IF(FLOCT(J))GO TO 450	QAP4	691
002845	000		KEEP=KEEP+1	QAP4	692
002846	000		INDEX(KEEP)=J	QAP4	693
002847	000	450	CONTINUE	QAP4	694
002848	000		WRITE(6,23)	QAP4	695
002849	000		WRITE(6,11)(INDEX(I),J=1,ICOL)	QAP4	696
002850	000		DO 460 I=1,IROW	QAP4	697
002851	000	460	WRITE(6,1)INDEX(I),(WORK1(I,J),J=1,ICOL),WORK2(I)	QAP4	698
002852	000		WRITE(6,7)	QAP4	699
002853	000	C	CALCULATE THE MODIFIED MATRIX AND GET COLUMN MINIMA	QAP4	700
002854	000	480	DO 500 J=1,ICOL	QAP4	701
002855	000		WORK3(J)=32767	QAP4	702
002856	000		DO 490 I=1,IROW	QAP4	703
002857	000		WORK1(I,J)=WORK1(I,J)-WORK3(J)	QAP4	704
002858	000		IF(WORK1(I,J).GF.WORK3(J))GO TO 490	QAP4	705
002859	000		WORK3(J)=WORK1(I,J)	QAP4	706
002860	000	490	CONTINUE	QAP4	707
002861	000	500	CONTINUE	QAP4	708
002862	000	C		QAP4	709
002863	000	C	PRINT COLUMN MINIMA	QAP4	710
002864	000		IF(KSNSW(23).EQ.0)GO TO 530	QAP4	711
002865	000		WRITE(6,11)(WORK3(J),J=1,ICOL)	QAP4	712
002866	000	C		QAP4	713
002867	000	C	IN CASE ICOL.GT.IROW, TAKE THE FIRST IROW COLUMNS AFTER RANKING	QAP4	714
002868	000	530	IF(ICOL.EQ.IROW)GO TO 540	QAP4	715
002869	000		CALL ARRNG(ICOL,1,1)	QAP4	716
002870	000	540	DO 550 J=1,IROW	QAP4	717
002871	000	550	IC2=IC2+WORK3(J)	QAP4	718
002872	000	C		QAP4	719
002873	000	C	WRITE THE VALUES OF IC2 IF YOU WISH	QAP4	720
002874	000		IF(KSNSW(24).EQ.1)WRITE(6,24)IC2	QAP4	721
002875	000	C		QAP4	722

002876	000	C		QAP4 723
002877	000	C	CALCULATE THE VALUE OF IC3	QAP4 724
002878	000	580	IC3=0	QAP4 725
002879	000		IF(IROW.EQ.1)GO TO 700	QAP4 726
002880	000		IF(KSNSW(26).EQ.1)WRITE(6,26)	QAP4 727
002881	000		KEEP=0	QAP4 728
002882	000		K=1	QAP4 729
002883	000		IROW2=IROW*(IROW-1)/2	QAP4 730
002884	000		DO 650 I=1,N2	QAP4 731
002885	000		IF(FLGDS(I))GO TO 650	QAP4 732
002886	000		DO 620 J=K,M2	QAP4 733
002887	000		IF(FLGIN(J))GO TO 620	QAP4 734
002888	000		KEEP=KEEP+1	QAP4 735
002889	000		II=DISTI(I)	QAP4 736
002890	000		IJ=DIJ(I)	QAP4 737
002891	000		JJ=INTRJ(J)	QAP4 738
002892	000		JJ=INTRJ(J)	QAP4 739
002893	000		KAM=D(II,IJ)*F(JI,JJ)	QAP4 740
002894	000		IC3=IC3+KAM	QAP4 741
002895	000		IF(KSNSW(26).EQ.1)WRITE(6,12)KEEP,II,IJ,JJ,D(II,IJ),F(JI,JJ)	QAP4 742
002896	000		*,KAM	QAP4 743
002897	000		GO TO 630	QAP4 744
002898	000	620	CONTINUE	QAP4 745
002899	000	630	IF(KEEP.EQ.IROW2)GO TO 700	QAP4 746
002900	000		K=J+1	QAP4 747
002901	000	650	CONTINUE	QAP4 748
002902	000		DO 660 I=1,10	QAP4 749
002903	000	660	WRITE(6,98)KEEP,IROW,IROW2	QAP4 750
002904	000		STOP	QAP4 751
002905	000	C		QAP4 752
002906	000	700	IF(KSNSW(27).EQ.1)WRITE(6,13)IC3	QAP4 753
002907	000		IF(MOVE.NE.0)GO TO 720	QAP4 754
002908	000		KEEP=IC1+IC2+IC3	QAP4 755
002909	000		IF(KEEP.GT.LBCST)LBCST=KEEP	QAP4 756
002910	000		WRITE(6,27)KEEP,LBCST	QAP4 757
002911	000		GO TO 750	QAP4 758
002912	000	720	LBCST=IC1+IC2+IC3	QAP4 759
002913	000	750	IF(KSNSW(28).EQ.1)WRITE(6,14)LBCST	QAP4 760
002914	000		IF(KSNSW(29).EQ.1)WRITE(6,17)IC1,IC2,IC3,LBCST	QAP4 761
002915	000	C		QAP4 762
002916	000	C		QAP4 763
002917	000		1 FORMAT(10X,20I5)	QAP4 764
002918	000		7 FORMAT(/)	QAP4 765
002919	000		11 FORMAT(15X,20I5/)	QAP4 766
002920	000		12 FORMAT(12X,I2,6X,I2,2X,I2,5X,I2,3X,I2,3(7X,I4))	QAP4 767
002921	000		14 FORMAT(90X,'VALUE OF C3 =',I5/)	QAP4 768
002922	000		16 FORMAT(90X,'LOWER BOUND =',I5/)	QAP4 769
002923	000		17 FORMAT(150X,'C1=',I5,5X,'C2=',I5,5X,'C3=',I5,10X,'LBCST=',I7)	QAP4 770
002924	000		18 FORMAT(1H0,' RESULTS FOR CALCULATIONS FOR LB AT THE START')	QAP4 771
002925	000		19 FORMAT(1H0,' RESULTS FOR CALCULATIONS FOR LB AT ITER ',I3,')	QAP4 772
002926	000		*LEVEL',I3,' ON A FORWARD MOVE')	QAP4 773
002927	000		21 FORMAT(1H0,' RESULTS FOR CALCULATIONS FOR LB AT ITER ',I3,')	QAP4 774
002928	000		*LEVEL',I3,' ON A BACKWARD MOVE')	QAP4 775
002929	000		22 FORMAT(90X,'VALUE IF C1 =',I5/)	QAP4 776
002930	000		23 FORMAT(10X,'MATRIX FOR CALCULATING C2 FOLLOWS'/)	QAP4 777
002931	000		24 FORMAT(90X,'VALUE OF C2 =',I5/)	QAP4 778
002932	000		26 FORMAT(10X,'ARRAY OF ELEMENTS TO CALCULATE C3 FOLLOWS'//,	QAP4 779

002933	000	* 10X, INDEX FROM 10 INTER BET. DIST. INTER VAL	GAP4 780
002934	000	*UE)	GAP4 781
002935	000	27 FORMAT(/10X,'CALCULATED STARTING LB=',I5,' SO WE TAKE LHCST=',	GAP4 782
002936	000	*I5)	GAP4 783
002937	000	98 FORMAT(02X,20(1H*),5(1) (LROW IN LWBND ==:)) KLEP='I3,' IROW=	GAP4 784
002938	000	*I,I3,' IROW2='I3)	GAP4 785
002939	000	RETURN	GAP4 786
002940	000	END	GAP4 787
002941	000	DEFLT,SI TUMER-S*SINAN-GAP4.MAX,,,285505053012	
002942	000	FUNCTION MAX(K,L)	GAP4 788
002943	000	MAX=K	GAP4 789
002944	000	IF(L.GT.K)MAX=L	GAP4 790
002945	000	RETURN	GAP4 791
002946	000	END	GAP4 792
002947	000	DELT,SI TUMER-S*SINAN-GAP4.MAX,,,27324053112	
002948	000	C A SUBROUTINE TO READ A QUADRATIC AND LINEAR PROBLEM	GAP4 793
002949	000	C THIS VERSION IS INTENDED FOR THE TRIPLET IT ALGORITHM	GAP4 794
002950	000	SUBROUTINE DTAIN(JOB,NFCIL,NFCCL,ITOT,ITOT,ITEM,LCST,ITOT)	GAP4 795
002951	000	COMMON IC1,IC2,IC3,IC4,IC5,IC6,IC7,IC8,IC9,IC10,IC11,IC12	GAP4 796
002952	000	COMMON MSTER(64000),IROW(1000),KNSNW(64)	GAP4 797
002953	000	COMMON KLPND,KUPND	GAP4 798
002954	000	INTEGER F(20,20),D(10,10),IX(20),IWORK(1000)	GAP4 799
002955	000	INTEGER WORK1(20,10)	GAP4 800
002956	000	EQUIVALENCE(MSTER(1),F(1,1))	GAP4 801
002957	000	EQUIVALENCE(MSTER(1000),F(10,1))	GAP4 802
002958	000	EQUIVALENCE(MSTER(10000),F(100,1))	GAP4 803
002959	000	EQUIVALENCE(MSTER(100000),F(1000,1))	GAP4 804
002960	000	EQUIVALENCE(IWORK(1000),F(1000,1))	GAP4 805
002961	000	C	GAP4 806
002962	000	C READ THE NUMBER OF FACILITIES, THE NUMBER OF LOCATIONS, THE FRACTION	GAP4 807
002963	000	C OF THE INTERVAL (ITOT,ITOT), THE NUMBER OF FIXED FACILITIES, AND	GAP4 808
002964	000	C THE CONTINUATION PARAMETER	GAP4 809
002965	000	READ(5,1)JOB,NFCIL,NFCCL,ITOT,ITOT,ITEM,LCST,ITOT	GAP4 810
002966	000	C	GAP4 811
002967	000	IF(ITOT.EQ.0)ITOT=3276700	GAP4 812
002968	000	DO 10 I=1,20	GAP4 813
002969	000	DO 10 J=1,20	GAP4 814
002970	000	F(I,J)=0	GAP4 815
002971	000	10 D(I,J)=0	GAP4 816
002972	000	C	GAP4 817
002973	000	DO 20 I=1,NFCIL	GAP4 818
002974	000	20 READ(5,2)(WORK1(I,J),J=1,NFCIL)	GAP4 819
002975	000	C	GAP4 820
002976	000	DO 30 I=1,NFCIL	GAP4 821
002977	000	DO 25 J=1,NFCIL	GAP4 822
002978	000	IF(I.EQ.J)GO TO 25	GAP4 823
002979	000	IF(I.GT.J)F(I,J)=WORK1(I,J)	GAP4 824
002980	000	IF(I.LT.J)D(I,J)=WORK1(I,J)	GAP4 825
002981	000	25 CONTINUE	GAP4 826
002982	000	30 CONTINUE	GAP4 827
002983	000	C	GAP4 828
002984	000	DO 40 I=1,NFCIL	GAP4 829
002985	000	DO 40 J=1,NFCIL	GAP4 830
002986	000	F(I,J)=F(J,I)	GAP4 831
002987	000	D(J,I)=D(I,J)	GAP4 832
002988	000	40 CONTINUE	GAP4 833
002989	000	C	GAP4 834

002990	000	DO 50 I=1,NFCIL	GAP4 835
002991	000	DO 45 J=1,NFCIL	GAP4 836
002992	000	IF(I.GT.J)F(I,J)=0	GAP4 837
002993	000	45 CONTINUE	GAP4 838
002994	000	50 CONTINUE	GAP4 839
002995	000	C	GAP4 840
002996	000	KSNSW(1)=0	GAP4 841
002997	000	KSNSW(4)=1	GAP4 842
002998	000	KSNSW(6)=1	GAP4 843
002999	000	KSNSW(07)=1	GAP4 844
003000	000	KSNSW(08)=1	GAP4 845
003001	000	C READ THE INDICES OF THE FACILITIES FIXED APRIORI	GAP4 846
003002	000	IF(NFIXD.F0.0)GO TO 00	GAP4 847
003003	000	DO 70 I=1,NFIXD	GAP4 848
003004	000	70 READ(5,1) FIX(I),WHERE(I)	GAP4 849
003005	000	C	GAP4 850
003006	000	80 RETURN	GAP4 851
003007	000	C	GAP4 852
003008	000	C	GAP4 853
003009	000	C	GAP4 854
003010	000	1 FORMAT(20X,12I5)	GAP4 855
003011	000	2 FORMAT(10X,35I2)	GAP4 856
003012	000		
003013	000	END	
003014	000	DELTA,SI TUMER-S*SINAN-GAP4.TIMEOUT,,,205745060512	
003015	000	SUBROUTINE TIMEOUT	
003016	000	COMMON IC1,IC2,IC3,NFCIL,IL0CT,M2,M2,LRCST,ITOT	
003017	000	COMMON MSTER(04000),TWOR(2000),KSNSW(60)	
003018	000	COMMON KLRND,KUPOT	
003019	000	K=ITIME(I,J)	
003020	000	T=I/5000.	
003021	000	WRITE(6,2000)T	
003022	000	2000 FORMAT(' TIME',F10.3,' SEC')	
003023	000	RETURN	
003024	000	END	
003025	000	DELTA,SI TUMER-S*SINAN-GAP7.HAYN,,,130663053112	
003026	000	COMMON IC1,IC2,IC3,NFCIL,IL0CT,M2,M2,LRCST,ITOT	GAP7 1
003027	000	COMMON MSTER(06200),TWOR(2000),KSNSW(60)	GAP7 2
003028	000	COMMON KLRND,KHADY,KUPOT,KUPOT	GAP7 3
003029	000	INTEGER*2 F(40,40),DIST(40,40),DISTV(40,40),INTRI(40,40)	GAP7 4
003030	000	INTEGER*2 INTVL(40,40),LEVEL(40),LOCAT(40),FCLTY(40),PTCST(40)	GAP7 5
003031	000	INTEGER*2 FIXED(40),WHERE(40,40),CMCST(40),FIX(40),WHERE(40)	GAP7 6
003032	000	INTEGER*2 IPASS(40),ICOST(100),ILOCT(100,40),FOUND(100)	GAP7 7
003033	000	INTEGER*2 JBAS(40,40),LIRND(40)	GAP7 8
003034	000	LOGICAL*1 FCFLG(40),FLCGI(40),FLGDS(40,40),FLGIN(40,40)	GAP7 9
003035	000	LOGICAL*1 BANED(40,40)	GAP7 10
003036	000	C	GAP7 11
003037	000	LOGICAL*1 WORK2(1220),WORK3(1220)	GAP7 12
003038	000	INTEGER*2 WORK3(0120),WORK4(0940),WORK7(0120)	GAP7 13
003039	000	C	GAP7 14
003040	000	EQUIVALENCE( MSTER(0001),F(1,1))	GAP7 11
003041	000	EQUIVALENCE( MSTER(0001),J(1,1))	GAP7 16
003042	000	C	GAP7 17
003043	000	EQUIVALENCE( MSTER(1601),DISTI(1))	GAP7 18
003044	000	EQUIVALENCE( MSTER(2401),DISTV(1))	GAP7 19
003045	000	EQUIVALENCE( MSTER(3201),INTRI(1))	GAP7 20
003046	000	EQUIVALENCE( MSTER(4001),INTVL(1))	GAP7 21

003047	000	EQUIVALENCE( MSTFR(1),MSTFR(1))	GAP7	22
003048	000	EQUIVALENCE( MSTFR(2),MSTFR(2))	GAP7	23
003049	000	EQUIVALENCE( MSTFR(3),MSTFR(3))	GAP7	24
003050	000	EQUIVALENCE( MSTFR(4),MSTFR(4))	GAP7	25
003051	000	EQUIVALENCE( MSTFR(5),MSTFR(5))	GAP7	26
003052	000	EQUIVALENCE( MSTFR(6),MSTFR(6))	GAP7	27
003053	000	EQUIVALENCE( MSTFR(7),MSTFR(7))	GAP7	28
003054	000	EQUIVALENCE( MSTFR(8),MSTFR(8))	GAP7	29
003055	000	EQUIVALENCE( MSTFR(9),MSTFR(9))	GAP7	30
003056	000	EQUIVALENCE( MSTFR(10),MSTFR(10))	GAP7	31
003057	000	EQUIVALENCE( MSTFR(11),MSTFR(11))	GAP7	32
003058	000	EQUIVALENCE( MSTFR(12),MSTFR(12))	GAP7	33
003059	000	EQUIVALENCE( MSTFR(13),MSTFR(13))	GAP7	34
003060	000	C	GAP7	35
003061	000	EQUIVALENCE( MSTFR(14),MSTFR(14))	GAP7	36
003062	000	EQUIVALENCE( MSTFR(15),MSTFR(15))	GAP7	37
003063	000	C	GAP7	38
003064	000	EQUIVALENCE( IWORK(1001),IWORK(1,1))	GAP7	39
003065	000	EQUIVALENCE( IWORK(1002),IWORK(1,1))	GAP7	40
003066	000	C	GAP7	41
003067	000	C	GAP7	42
003068	000	C	GAP7	43
003069	000	READ THE KEYS SETTINGS FOR THE MAIN PROGRAM	GAP7	44
003070	000	READ(5,3)(KSNSW(I),I=1,70)	GAP7	45
003071	000	C	GAP7	46
003072	000	READ THE KEY SETTINGS FOR THE ROUTINE LWBNDP	GAP7	47
003073	000	READ(5,3)(KSNSW(I),I=1,36)	GAP7	48
003074	000	C	GAP7	49
003075	000	READ THE PROBLEM DATA	GAP7	50
003076	000	10 CALL DTAIN(JOB,IFRAC,ISDB,IMDE,ITERM)	GAP7	51
003077	000	C	GAP7	52
003078	000	PRINT THE INITIAL DATA OF THE PROBLEM	GAP7	53
003079	000	IF(KSNSW(01).EQ.0)GO TO 120	GAP7	54
003080	000	WRITE(6,4)JOB,NFCIL,NLOCT	GAP7	55
003081	000	DO 70 I=1,2	GAP7	56
003082	000	70 WRITE(6,2)	GAP7	57
003083	000	WRITE(6,5)	GAP7	58
003084	000	DO 80 I=1,NLOCT	GAP7	59
003085	000	80 WRITE(6,01)(D(I,J),J=1,NLOCT)	GAP7	60
003086	000	WRITE(6,8)	GAP7	61
003087	000	DO 90 I=1,NFCIL	GAP7	62
003088	000	90 WRITE(6,01)(F(I,J),J=1,NFCIL)	GAP7	63
003089	000	WRITE(6,07)	GAP7	64
003090	000	WRITE(6,5)I,BCST,ITOT	GAP7	65
003091	000	C	GAP7	66
003092	000	INITIALIZE THE COST	GAP7	67
003093	000	110 IC1=0	GAP7	68
003094	000	IF(NFIXD.EQ.1)GO TO 125	GAP7	69
003095	000	DO 120 I=1,10	GAP7	70
003096	000	120 WRITE(6,98)NFXD	GAP7	71
003097	000	GO TO 900	GAP7	72
003098	000	C	GAP7	73
003099	000	UNFLAG ALL THE LOCATIONS AND THE FACILITIES	GAP7	74
003100	000	125 DO 130 I=1,NLOCT	GAP7	75
003101	000	130 FLOGT(I)=.FALSE.	GAP7	76
003102	000	DO 140 J=1,NFCIL	GAP7	77
003103	000	140 FCFLG(J)=.FALSE.	GAP7	78
		DO 150 I=1,NFCIL		
		IPASS(I)=0		
		DO 150 J=1,NLOCT		

003104	000	150 BANE(I,J)=.FALSE.	GAP7	79
003105	000	C	GAP7	80
003106	000	C INITIALIZE WORK7	GAP7	81
003107	000	DO 325 I=1,120	GAP7	82
003108	000	325 WORK7(I)=0	GAP7	83
003109	000	C INITIALIZE WORK8	GAP7	84
003110	000	DO 330 I=1,1220	GAP7	85
003111	000	330 WORK8(I)=.FALSE.	GAP7	86
003112	000	DO 335 I=1,40	GAP7	87
003113	000	335 CMCST(I)=0	GAP7	88
003114	000	C	GAP7	89
003115	000	C	GAP7	90
003116	000	C	GAP7	91
003117	000	C	GAP7	92
003118	000	TIME=0	GAP7	93
003119	000	IEVAL=0	GAP7	94
003120	000	IUBCS=0	GAP7	95
003121	000	KOMPL=0	GAP7	96
003122	000	FRAC=IFRAC	GAP7	97
003123	000	FRAC=0.01*FRAC	GAP7	98
003124	000	C	GAP7	99
003125	000	C INITIALIZE SUBROUTINE COUNTING AND FATHOMING COUNTER	GAP7	100
003126	000	KLBND=0	GAP7	101
003127	000	KHADY=0	GAP7	102
003128	000	KCARY=0	GAP7	103
003129	000	KEATH=0	GAP7	104
003130	000	NECTP=0	GAP7	105
003131	000	C	GAP7	106
003132	000	C CALCULATE STARTING UPPER AND LOWER BOUNDS	GAP7	107
003133	000	CALL LWOND2(0,0,0,0,1)	GAP7	108
003134	000	LBTC=LRCST	GAP7	109
003135	000	IF(KSNSW(04).EQ.1)WRITE(6,9)LBTC,ITOT	GAP7	110
003136	000	C	GAP7	111
003137	000	C PROCEED TO OBTAIN A SOLUTION ALONG THE RIGHTMOST BRANCH OF A TREE	GAP7	112
003138	000	ITER=1	GAP7	113
003139	000	LEVY=1	GAP7	114
003140	000	L2= FIX(1)	GAP7	115
003141	000	J=WHERE(1)	GAP7	116
003142	000	LEVEL(L2)=LEVY	GAP7	117
003143	000	LOCAT(L2)=J	GAP7	118
003144	000	FCLG(L2)=.TRUE.	GAP7	119
003145	000	FLOGT(J)=.TRUE.	GAP7	120
003146	000	IF(KSNSW(05).EQ.1)WRITE(6,9)(ITER,LEVY,L2,J	GAP7	121
003147	000	PTCST(1)=0	GAP7	122
003148	000	CMCST(1)=0	GAP7	123
003149	000	C	GAP7	124
003150	000	J=FIX(1)	GAP7	125
003151	000	FCLTY(1)=J	GAP7	126
003152	000	LEVEL(J)=1	GAP7	127
003153	000	FIXED(1)=J	GAP7	128
003154	000	LIXED(1)=WHERE(1)	GAP7	129
003155	000	C B) CHOOSE THE ELEMENT WITH MAXIMUM INTERACTIONS WITH THE ONE ASSIGNED	GAP7	130
003156	000	C AT THE PRECEDING LEVEL	GAP7	131
003157	000	430 LEVY=LEVY+1	GAP7	132
003158	000	IPASS(LEVY)=IPASS(LEVY)+1	GAP7	133
003159	000	ITER=ITER+1	GAP7	134
003160	000	JET=FCLTY(LEVY-1)	GAP7	135

003161	000	ISAV=-32767	0AP7 136
003162	000	DO 460 I=1,NFCIL	0AP7 137
003163	000	IF(FCFLG(I))GO TO 460	0AP7 138
003164	000	IF(F(I,JET).LE.ISAV)GO TO 460	0AP7 139
003165	000	ISAV=F(I,JET)	0AP7 140
003166	000	KEEP=I	0AP7 141
003167	000	460 CONTINUE	0AP7 142
003168	000	470 FCFLG(KEEP)=.TRUE.	0AP7 143
003169	000	LEVEL(KEEP)=LEVY	0AP7 144
003170	000	C C) ASSIGN THE ELEMENT TO A LOCATION SO THAT THE COST OF SUCH AN ASSI	0AP7 145
003171	000	C IS MINIMAL	0AP7 146
003172	000	510 IWO=32767	0AP7 147
003173	000	DO 530 J=1,NLOC	0AP7 148
003174	000	IF(UBAND(KEEP,J))GO TO 530	0AP7 149
003175	000	IF(FLOGT(J))GO TO 530	0AP7 150
003176	000	KAM=0	0AP7 151
003177	000	DO 520 I=1,NFCIL	0AP7 152
003178	000	IF(I.LE.KEEP)GO TO 520	0AP7 153
003179	000	IF(.NOT.FCFLG(I))GO TO 520	0AP7 154
003180	000	L=LOCAT(I)	0AP7 155
003181	000	ISAVV=F(KEEP,I)	0AP7 156
003182	000	KAM=KAM+(ISAVV*D(J,L))	0AP7 157
003183	000	520 CONTINUE	0AP7 158
003184	000	IF(KAM.GT.IWD)GO TO 530	0AP7 159
003185	000	IWO=KAM	0AP7 160
003186	000	LOCAT(KEEP)=J	0AP7 161
003187	000	530 CONTINUE	0AP7 162
003188	000	IF(IWO.EQ.32767)GO TO 730	0AP7 163
003189	000	IF(KSNOW(05).EQ.1)WRITE(6,9)ITER,LEVY,KEEP,LOCAT(KEEP),IWD	0AP7 164
003190	000	J=LOCAT(KEEP)	0AP7 165
003191	000	FCLTY(LEVY)=KEEP	0AP7 166
003192	000	FLOGT(J)=.TRUE.	0AP7 167
003193	000	IUBCS=IUBCS+IWO	0AP7 168
003194	000	FIXED(LEVY)=KEEP	0AP7 169
003195	000	PTCST(LEVY)=IWD	0AP7 170
003196	000	CHCST(LEVY)=IUBCS	0AP7 171
003197	000	LIXED(LEVY)=J	0AP7 172
003198	000	C	0AP7 173
003199	000	C CHECK WHETHER WE HAVE REACHED A TERMINAL NODE OR NOT	0AP7 174
003200	000	580 IF(LEVY.LT.NFCIL.AND.KFATH.LE.0)GO TO 430	0AP7 175
003201	000	IF(LEVY.EQ.NFCIL)GO TO 590	0AP7 176
003202	000	C CHECK WHETHER IT PAYS TO CONTIN THE SEARCH ALONG THIS BRANCH	0AP7 177
003203	000	CALL LWBND2(LEVY,ITER,IWD,1,2)	0AP7 178
003204	000	IF(LBCST.LT.ITOT)GO TO 540	0AP7 179
003205	000	KFATH=KFATH+1	0AP7 180
003206	000	GO TO 620	0AP7 181
003207	000	585 LEVY=LEVY+1	0AP7 182
003208	000	IPASS(LEVY)=IPASS(LEVY)+1	0AP7 183
003209	000	GO TO 730	0AP7 184
003210	000	C E) NOW WE HAVE A COMPLETE SOLUTION , PRINT IT IF YOU WISH	0AP7 185
003211	000	590 KOMPL=1	0AP7 186
003212	000	IC1=IUBCS	0AP7 187
003213	000	IF(IUBCS.GE.ITOT)GO TO 620	0AP7 188
003214	000	ITOT=IUBCS	0AP7 189
003215	000	IEVAL=IEVAL+1	0AP7 190
003216	000	ICOST(IEVAL)=IUBCS	0AP7 191
003217	000	FOUND(IEVAL)=ITER	0AP7 192



003218	000	DO 595 I=1,NFCIL	GAP7 193
003219	000	ILOCT(IEVAL,I)=LOCAT(I)	GAP7 194
003220	000	595 CONTINUE	GAP7 195
003221	000	600 IF(KSNSW(06).EQ.0)GO TO 620	GAP7 196
003222	000	WRITE(6,11)ITER	GAP7 197
003223	000	WRITE(6,12)	GAP7 198
003224	000	DO 605 I=1,LEVY	GAP7 199
003225	000	605 WRITE(6,14)I,FIXED(I),LIXED(I),PTCST(I),CMCST(I)	GAP7 200
003226	000	CALL TIMEOUT	
003227	000	C	GAP7 201
003228	000	C REDUCE THE CURRENT UPPER BOUND	GAP7 202
003229	000	610 DIF=ITOT-LBTC	GAP7 203
003230	000	KETOT=ITOT	GAP7 204
003231	000	KELUC=LBTC	GAP7 205
003232	000	IF(DIF.LE.ISTOP)GO TO 640	GAP7 206
003233	000	DIF=(FRAC*DIF)	GAP7 207
003234	000	ITOT=LBTC+DIF	GAP7 208
003235	000	IF(KSNSW(08).EQ.1)WRITE(6,13)KELUC,ITOT,KETOT	GAP7 209
003236	000	C STORE THE TREE CONFIGURATION AT THE CURRENT NODE SO THAT WE CAN	GAP7 210
003237	000	C RETRIEVE IT IF WE FAIL TO OBTAIN A BETTER SOLUTION UNDER THE CURRENT	GAP7 211
003238	000	C CEILING	GAP7 212
003239	000	DO 612 I=1,1220	GAP7 213
003240	000	WORK2(I)=.FALSE.	GAP7 214
003241	000	IF(WORK8(I))WORK2(I)=.TRUE.	GAP7 215
003242	000	612 CONTINUE	GAP7 216
003243	000	DO 615 I=1,120	GAP7 217
003244	000	615 WORK3(I)=WORK7(I)	GAP7 218
003245	000	DO 616 I=1,40	GAP7 219
003246	000	616 WORK6(I)=CMCST(I)	GAP7 220
003247	000	CALL TIMEOUT	
003248	000	C	GAP7 221
003249	000	C *** BACKTRACK TO THE NEXT LEVEL UP. THE TREE	GAP7 222
003250	000	C A) FREE THE FACILITY AT THAT LEVEL	GAP7 223
003251	000	620 DO 640 I=1,NFCIL	GAP7 224
003252	000	IF(LEVEL(I).EQ.LEVY)GO TO 650	GAP7 225
003253	000	640 CONTINUE	GAP7 226
003254	000	C	GAP7 227
003255	000	650 FCFLG(I)=.FALSE.	GAP7 228
003256	000	DO 670 J=1,NLOCT	GAP7 229
003257	000	IF(J.EQ.LOCAT(I))GO TO 680	GAP7 230
003258	000	670 CONTINUE	GAP7 231
003259	000	680 FLOGT(J)=.FALSE.	GAP7 232
003260	000	IWD=-PTCST(LEVY)	GAP7 233
003261	000	IUBCS=IUBCS+IWD	GAP7 234
003262	000	BANED(I,J)=.TRUE.	GAP7 235
003263	000	C	GAP7 236
003264	000	C	GAP7 237
003265	000	C	GAP7 238
003266	000	C C) CALCULATE A LB ON ALL SOLUTIONS EXCLUDING THE LAST	GAP7 239
003267	000	CALL LWBND2(LEVY,ITER,IWD,2,2)	GAP7 240
003268	000	C	GAP7 241
003269	000	C D) CUECK WHETHER IT PAYS TO RESUME THE SEARCH ALONG THIS BRANCH	GAP7 242
003270	000	IF(LBCST.LT.ITOT)GO TO 730	GAP7 243
003271	000	KFATH=KFATH+1	GAP7 244
003272	000	KEEP=I	GAP7 245
003273	000	GO TO 740	GAP7 246
003274	000	C	GAP7 247

003275	000	C	RETRIEVE THE INDEX OF THE FACILITY AT THAT LEVEL	GAP7 248
003276	000	730	KEEP=FIXED(LEVY)	GAP7 249
003277	000		ITER=ITER+1	GAP7 250
003278	000		GO TO 470	GAP7 251
003279	000	C		GAP7 252
003280	000	740	DO 760 J=1,NLOC	GAP7 253
003281	000	760	BANED(KEEP,J)=.FALSE.	GAP7 254
003282	000		FCFLG(KEEP)=.FALSE.	GAP7 255
003283	000		LEVY=LEVY-1	GAP7 256
003284	000		ITER=ITER+1	GAP7 257
003285	000		IF(LEVY.NE.0.AND.ITER.NE.ITER0)GO TO 690	GAP7 258
003286	000		IF(LEVY.EQ.0.AND.ITER.LT.ITER0)GO TO 775	GAP7 259
003287	000		WRITE(6,6)	GAP7 260
003288	000		WRITE(6,39)LRIC,ITOT	GAP7 261
003289	000		WRITE(6,7)	GAP7 262
003290	000		DO 765 I=1,2	GAP7 263
003291	000	765	WR ITE(6,41)ITER	GAP7 264
003292	000		GO TO 810	GAP7 265
003293	000	C	SINCE NO BETTER SOLUTION WAS FOUND , FLIP THE BOUNDS , RESTORE THE	GAP7 266
003294	000	C	ARRAYS AND RETURN TO 690	GAP7 267
003295	000	775	IF(KSNSW(08).EQ.1)WRITE(6,37)LRIC,ITOT	GAP7 268
003296	000		IF(KSNSW(08).EQ.1)WRITE(6,38)ITER	GAP7 269
003297	000		CALL TIMOUT	GAP7 270
003298	000		LRIC=ITOT	GAP7 271
003299	000		KFLBC=LRIC	GAP7 272
003300	000		ICI=KETOT	GAP7 273
003301	000		IUCCS=KETOT	GAP7 274
003302	000		ITOT=KETOT	GAP7 275
003303	000		DIF=ITOT-LRIC	GAP7 276
003304	000		IF(DIF.LT.1STOP)GO TO 778	GAP7 277
003305	000		DIF=FRAC*DIF	
003306	000		GO TO 780	
003307	000	778	IF(KSNSW(08).EQ.1)WRITE(6,37)LRIC	GAP7 278
003308	000		GO TO 800	GAP7 279
003309	000	C		GAP7 280
003310	000	780	KETOT=ITOT	GAP7 281
003311	000		ITOT=LRIC+DIF	GAP7 282
003312	000		DO 785 I=1,1220	GAP7 283
003313	000		WORK8(I)=.FALSE.	GAP7 284
003314	000		IF(WORK2(I))WORK8(I)=.TRUE.	GAP7 285
003315	000	785	CONTINUE	GAP7 286
003316	000		DO 790 I=1,120	GAP7 287
003317	000	790	WORK7(I)=WORK3(I)	GAP7 288
003318	000		DO 795 I=1,40	GAP7 289
003319	000	795	CMCST(I)=WORK6(I)	GAP7 290
003320	000		LEVY=NFCIL	GAP7 291
003321	000	C		GAP7 292
003322	000		IF(KSNSW(08).EQ.1)WRITE(6,38)LRIC,ITOT	GAP7 293
003323	000		GO TO 620	GAP7 294
003324	000	C		GAP7 295
003325	000	C		GAP7 296
003326	000	C		GAP7 297
003327	000	800	WRITE(6,16)	GAP7 298
003328	000	810	KEEP=0	GAP7 299
003329	000		FTHRIO=100.0*FLOAT(KFATH)/FLOAT(KLBND)	GAP7 300
003330	000		DO 820 I=1,NFCIL	GAP7 301
003331	000		LIXED(I)=I	GAP7 302

003332	000	820	KEEP=KEEP+IPASS(I)		QAP7 303
003333	000		WRITE(6,23)		QAP7 304
003334	000		WRITE(6,24)NFCIL		QAP7 305
003335	000		WRITE(6,26)NLOCT		QAP7 306
003336	000		WRITE(6,07)		QAP7 307
003337	000		WRITE(6,47)NFI XO		QAP7 308
003338	000		WRITE(6,48)		QAP7 309
003339	000		DO 825 I=1,NFI XO		QAP7 310
003340	000	825	WRITE(6,49)I,FI X(I),WHERE(I)		QAP7 311
003341	000		WRITE(6,08)		QAP7 312
003342	000		DO 830 I=1,NFCIL		QAP7 313
003343	000	830	WRITE(6,1)(F(I,J),J=1,NFCIL)		QAP7 314
003344	000		WRITE(6,05)		QAP7 315
003345	000		DO 840 I=1,NLOCT		QAP7 316
003346	000	840	WRITE(6,1)(D(I,J),J=1,NLOCT)		QAP7 317
003347	000		WRITE(6,27)IILR		QAP7 318
003348	000		WRITE(6,28)(IPASS(I),I=1,NFCIL)		QAP7 319
003349	000		WRITE(6,29)KEEP		QAP7 320
003350	000		WRITE(6,31)		QAP7 321
003351	000		WRITE(6,32)(LIXED(I),I=1,NFCIL)		QAP7 322
003352	000		DO 850 I=1,IEVAL		QAP7 323
003353	000	850	WRITE(6,33)I,ICOST(I),FOUN D(I),(ILOCT(I,L),L=1,NFCIL)		QAP7 324
003354	000		WRITE(6,34)TIME		QAP7 325
003355	000		WRITE(6,07)		QAP7 326
003356	000		WRITE(6,42)		QAP7 327
003357	000		WRITE(6,43)KLBND		QAP7 328
003358	000		WRITE(6,44)NECTP		QAP7 329
003359	000		WRITE(6,52)KHADY		QAP7 330
003360	000		WRITE(6,53)KGARY		QAP7 331
003361	000		WRITE(6,46)FTHRIO		QAP7 332
003362	000		WRITE(6,06)		QAP7 333
003363	000	C			QAP7 334
003364	000	C			QAP7 335
003365	000	C	CHECK WHETHER THERE ARE ANY MORE PROBLEMS TO SOLVE		QAP7 336
003366	000		900 IF(MORE)1000,1000,10		QAP7 337
003367	000	C			QAP7 338
003368	000		1000 STOP		QAP7 339
003369	000	C			QAP7 340
003370	000		1 FORMAT(10X,40I3)		QAP7 341
003371	000		2 FORMAT(10X,9('*** QAP 7 -'))		QAP7 342
003372	000		3 FORMAT(20X,20I3)		QAP7 343
003373	000		4 FORMAT(1H1,' DATA FOR PROBLEM',I5/		QAP7 344
003374	000		* 10X,' NUMBER OF FACILITIES =',I3/		QAP7 345
003375	000		* 10X,' NUMBER OF LOCATIONS =',I3//)		QAP7 346
003376	000		5 FORMAT(10X,' DISTANCE MATRIX//)		QAP7 347
003377	000		6 FORMAT(1H1)		QAP7 348
003378	000		7 FORMAT(/)		QAP7 349
003379	000		8 FORMAT(/10X,'INTERACTIONS MATRIX//)		QAP7 350
003380	000		9 FORMAT(20X,'AT ITERATION ',I4,' LEVEL',I3,' FACILITY',I3,' IS		QAP7 351
003381	000		*LOCATED',I3,' WITH AN INCREASE IN COST LEVELS',I5)		QAP7 352
003382	000		11 FORMAT(1H0,5X,'RESULTS OF THE MOST RECENT FEASIBLE SOLUTION , FOUN		QAP7 353
003383	000		*D AT ITERATION',I6//)		QAP7 354
003384	000		12 FORMAT(20X,'LEVEL FACILITY LOCATION PI CST CMCST//)		QAP7 355
003385	000		14 FORMAT(22X,I2,10X,I2,0X,I3,0',I4,4X,I5)		QAP7 356
003386	000		16 FORMAT(10X,'END OF CALCULATIONS')		QAP7 357
003387	000		17 FORMAT(/10X,'STARTING LOWER BOUND ON THE TOTAL COST=',I5/		QAP7 358
003388	000		* 10X,'STARTING UPPER BOUND ON THE TOTAL COST=',I5//)		QAP7 359

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003389 000 21 FORMAT(10X,' ARRAY OF VARIOUS DISTANCE VALUES') GAP7 360
003390 000 22 FORMAT(10X,' ARRAY OF VARIOUS INTERACTION VALUES') GAP7 361
003391 000 23 FORMAT(1H,' SUMMARY OF QUADRATIC ASSIGNMENT PROBLEM') GAP7 362
003392 000 24 FORMAT(10X,' NUMBER OF ELEMENTS =',I5) GAP7 363
003393 000 25 FORMAT(10X,' NUMBER OF LOCATIONS =',I5) GAP7 364
003394 000 26 FORMAT(10X,' NUMBER OF VARIOUS LOCATIONS =',I5) GAP7 365
003395 000 27 FORMAT(10X,' NUMBER OF VARIOUS LOCATIONS =',I5) GAP7 366
003396 000 28 FORMAT(10X,' NUMBER OF VARIOUS LOCATIONS =',I5) GAP7 367
003397 000 29 FORMAT(10X,' TOTAL NUMBER OF LOCATIONS =',I5) GAP7 368
003398 000 30 FORMAT(10X,' NUMBER OF LOCATIONS =',I5) GAP7 369
003399 000 31 FORMAT(10X,' TOTAL NUMBER OF LOCATIONS =',I5) GAP7 370
003400 000 32 FORMAT(10X,' TOTAL NUMBER OF LOCATIONS =',I5) GAP7 371
003401 000 33 FORMAT(10X,' TOTAL NUMBER OF LOCATIONS =',I5) GAP7 372
003402 000 C 34 FORMAT(10X,' A SUMMARY OF THIS NEWLY FORMED SOLUTION , WE HAVE , GAP7 373
003403 000 *LATEST LB=,I5,' *LATEST LB=,I5,' *LATEST LB=,I5) GAP7 374
003404 000 35 FORMAT(10X,' HOWEVER , IT DOES NOT SEEM TO BE WORTH IT AS DIF=, GAP7 375
003405 000 *1E20.8) GAP7 376
003406 000 36 FORMAT(15X,' THE CURRENT LOWER AND UPPER BOUNDS ARE ',I17) GAP7 377
003407 000 37 FORMAT(10X,' SINCE THE CURRENT SOLUTION HAS BEEN BETWEEN',I5,' + ',I5,' GAP7 378
003408 000 *I5,' I WILL RESTART AT THE LAST COMPLETE SOLUTION) GAP7 379
003409 000 38 FORMAT(10X,' I WILL RESTART AT THE LAST COMPLETE SOLUTION) GAP7 380
003410 000 39 FORMAT(10X,' I WILL RESTART AT THE LAST COMPLETE SOLUTION) GAP7 381
003411 000 40 FORMAT(10X,' I WILL RESTART AT THE LAST COMPLETE SOLUTION) GAP7 382
003412 000 C 41 FORMAT(10X,' NUMBER OF DATA FOR VARIOUS PARAMETERS') GAP7 383
003413 000 42 FORMAT(15X,' SUBROUTINE DTAIN WAS CALLED',I5,' TIMES') GAP7 384
003414 000 43 FORMAT(15X,' SUBROUTINE DTAIN WAS CALLED',I5,' TIMES') GAP7 385
003415 000 44 FORMAT(15X,' SUBROUTINE DTAIN WAS CALLED',I5,' TIMES') GAP7 386
003416 000 45 FORMAT(15X,' SUBROUTINE DTAIN WAS CALLED',I5,' TIMES') GAP7 387
003417 000 46 FORMAT(10X,' FATHOMING ERROR =',E18.4,' PER CENT') GAP7 388
003418 000 C 47 FORMAT(10X,' NUMBER OF DATA ELEMENTS',I5,' ARRANGED AS FOLLOWS' GAP7 389
003419 000 *') GAP7 390
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003434 000 C A SUBROUTINE TO READ DATA FOR THE QUADRATIC ASSIGNMENT PROBLEM GAP71100
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003440	000	C	READ THE NUMBER OF FACILITIES, THE NUMBER OF LOCATIONS, THE FRACTION	GAP71113
003447	000	C	OF THE INTERVAL (ITOT-1), THE NUMBER OF FIXED FACILITIES, AND	GAP71114
003448	000	C	THE CONTINUATION PARAMETER	GAP71115
003449	000		READ(5,1)JOB,NFCIL,NLOC,IFRAC,NFIXD,MORE,ISTOP,ITERM,LBCST,ITOT	GAP71116
003450	000	C		GAP71117
003451	000		IF(ITOT.EQ.0)ITOT=32767	GAP71118
003452	000		DO 10 I=1,40	GAP71119
003453	000		DO 10 J=1,40	GAP71120
003454	000		F(I,J)=0	GAP71121
003455	000		10 D(I,J)=0	GAP71122
003456	000	C		GAP71123
003457	000		DO 20 I=1,NFCIL	GAP71124
003458	000		20 READ(5,2)(WORK1(I,J),J=1,NFCIL)	GAP71125
003459	000	C		GAP71126
003460	000		DO 30 I=1,NFCIL	GAP71127
003461	000		DO 25 J=1,NFCIL	GAP71128
003462	000		IF(I.EQ.J)GO TO 25	GAP71129
003463	000		IF(I.GT.J)F(I,J)=WORK1(I,J)	GAP71130
003464	000		IF(I.LT.J)D(I,J)=WORK1(I,J)	GAP71131
003465	000		25 CONTINUE	GAP71132
003466	000		30 CONTINUE	GAP71133
003467	000	C		GAP71134
003468	000		DO 40 I=1,NFCIL	GAP71135
003469	000		DO 40 J=1,NFCIL	GAP71136
003470	000		F(I,J)=F(J,I)	GAP71137
003471	000		D(J,I)=D(I,J)	GAP71138
003472	000		40 CONTINUE	GAP71139
003473	000	C		GAP71140
003474	000	C		GAP71141
003475	000		KSNSW(1)=0	GAP71142
003476	000		KSNSW(4)=1	GAP71143
003477	000		KSNSW(6)=1	GAP71144
003478	000		KSNSW(07)=1	GAP71145
003479	000		KSNSW(08)=1	GAP71146
003480	000	C	READ THE INDICES OF THE FACILITIES FIXED A PRIORI	GAP71147
003481	000		DO 70 I=1,NFIXD	GAP71148
003482	000		70 READ(5,1) FIX(I),WHERE(I)	GAP71149
003483	000	C		GAP71150
003484	000		NFIXD=1	GAP71151
003485	000		RETURN	GAP71152
003486	000	C		GAP71153
003487	000	C		GAP71154
003488	000	C		GAP71155
003489	000		1 FORMAT(20X,12I5)	GAP71156
003490	000		2 FORMAT(10X,35I2)	GAP71157
003491	000			
003492	000		END	
003493	000		QFLT,SI TUMER-S*SINAN-QAP7.GARY,,,206344053012	
003494	000	C		GAP7 899
003495	000	C		GAP7 903
003496	000	C		GAP7 904
003497	000	C	*****	GAP7 905
003498	000	C	A SUBROUTINE TO IDENTIFY BASIC LOOP AND UPDATE ITS ENTRIES	GAP7 906
003499	000	C	<<*****	GAP7 907
003500	000		SUBROUTINE GARY(NORTG,K1,NROW1,NCOL1,KEY)	GAP7 908
003501	000	C	-----	GAP7 909
003502	000		COMMON IC1,IC2,IC3,NFCIL,NLOC,IFRAC,NFIXD,LBCST,KTOT	GAP7 910

003503	000	COMMON MSTER(002001,10000000),KSN5W(00)	GAP7 911
003504	000	COMMON KLRND,KHADR,KGARY,KGARY1	GAP7 912
003505	000	C	GAP7 913
003506	000	INTEGER*2 JBAS(100,100)	GAP7 914
003507	000	INTEGER*2 INET(100,100),INET2(50)	GAP7 915
003508	000	C	GAP7 916
003509	000	EQUIVALENCE( IWORK(100), JBAS(100,1))	GAP7 917
003510	000	EQUIVALENCE( IWORK(100), JBAS(100,1))	GAP7 918
003511	000	EQUIVALENCE( IWORK(100), JBAS(100,1))	GAP7 919
003512	000	EQUIVALENCE( IWORK(100), JBAS(5,1))	GAP7 920
003513	000	C	GAP7 921
003514	000	C	GAP7 922
003515	000	C	GAP7 923
003516	000	C	GAP7 924
003517	000	C	GAP7 925
003518	000	C	GAP7 926
003519	000	UPDATE CALLS COUNTER	GAP7 927
003520	000	KGARY=KGARY+1	GAP7 928
003521	000	C	GAP7 929
003522	000	A) INITIALIZE	GAP7 930
003523	000	* SET K2 .GE. (10*K)	GAP7 931
003524	000	K2=300000	GAP7 932
003525	000	K=NORIG+NDEST	GAP7 933
003526	000	K=2*K	GAP7 934
003527	000	DO 20 I=1,K	GAP7 935
003528	000	20 INET(I)=0	GAP7 936
003529	000	C	GAP7 937
003530	000	** THUS THE SIZE OF ARRAY INET .EQ. 2*(NORIG+NDEST) **	GAP7 938
003531	000	DO 40 I=1,NORIG	GAP7 939
003532	000	40 INET1(I)=0	GAP7 940
003533	000	DO 60 I=1,NDEST	GAP7 941
003534	000	60 INET2(I)=0	GAP7 942
003535	000	C	GAP7 943
003536	000	C	GAP7 944
003537	000	C	GAP7 945
003538	000	C	GAP7 946
003539	000	C	GAP7 947
003540	000	C	GAP7 948
003541	000	I=1	GAP7 949
003542	000	INET(I)=NROW1	GAP7 950
003543	000	I1=I+1	GAP7 951
003544	000	INET(I1)=NCOL1	GAP7 952
003545	000	NROW=NROW1	GAP7 953
003546	000	NCOL=1	GAP7 954
003547	000	I=I+2	GAP7 955
003548	000	100 IF(JBAS(NROW,NCOL)-R1)150,120,120	GAP7 956
003549	000	120 IF(NCOL-NCOL1)140,160,180	GAP7 957
003550	000	140 INET(I)=NROW	GAP7 958
003551	000	I1=I+1	GAP7 959
003552	000	INET(I1)=NCOL	GAP7 960
003553	000	I=I+2	GAP7 961
003554	000	GO TO 220	GAP7 962
003555	000	C	GAP7 963
003556	000	160 NCOL=NCOL+1	GAP7 964
003557	000	IF(NCOL-NDEST)100,100,200	GAP7 965
003558	000	200 I=1	GAP7 966
003559	000	C	GAP7 967
		*) NO BASIS ELEMENT IN THE ROW OF THE ENTERING , ERROR TYPE 1	

003560	000	WRITE(6,1)I,NROW1,NCOL1	QAP7 968
003561	000	KEY=2	QAP7 969
003562	000	C	QAP7 970
003563	000	RETURN	QAP7 971
003564	000	C <<+++++>>	QAP7 972
003565	000	240 INET2(NCOL)=1	QAP7 973
003566	000	NROW=1	QAP7 974
003567	000	240 IF(JBAS(NROW,NCOL)-K1)260,300,380	QAP7 975
003568	000	260 NROW=NROW+1	QAP7 976
003569	000	IF(NROW-NORIG)240,240,300	QAP7 977
003570	000	300 I=I-2	QAP7 978
003571	000	IF(I)360,360,340	QAP7 979
003572	000	340 NROW=INET(I)	QAP7 980
003573	000	I1=I+1	QAP7 981
003574	000	NCOL=INET(I1)	QAP7 982
003575	000	INET2(NCOL)=0	QAP7 983
003576	000	C	QAP7 984
003577	000	GO TO 520	QAP7 985
003578	000	C *) THERE IS NO BASIS LOOP	QAP7 986
003579	000	360 I=2	QAP7 987
003580	000	C	QAP7 988
003581	000	WRITE(6,2)I,NROW1,NCOL1	QAP7 989
003582	000	KEY=2	QAP7 990
003583	000	C	QAP7 991
003584	000	RETURN	QAP7 992
003585	000	C <<+++++>>	QAP7 993
003586	000	380 I2=I-2	QAP7 994
003587	000	IF(NROW-INET(I2))400,260,400	QAP7 995
003588	000	400 IF(INET1(NROW))300,420,500	QAP7 996
003589	000	420 INET(I)=NROW	QAP7 997
003590	000	I1=I+1	QAP7 998
003591	000	INET(I1)=NCOL	QAP7 999
003592	000	I=I+2	QAP71000
003593	000	IF(NROW-NROW1)480,460,400	QAP71001
003594	000	460 I=I-2	QAP71002
003595	000	GO TO 300	QAP71003
003596	000	460 INET1(NROW)=1	QAP71004
003597	000	NCOL=1	QAP71005
003598	000	500 IF(JBAS(NROW,NCOL)-K1)520,560,560	QAP71006
003599	000	520 NCOL=NCOL+1	QAP71007
003600	000	IF(NCOL-NDEST)500,500,600	QAP71008
003601	000	560 I1=I-1	QAP71009
003602	000	IF(NCOL-INET(I1))580,520,580	QAP71010
003603	000	580 IF(INET2(NCOL))640,600,640	QAP71011
003604	000	600 INET(I)=NROW	QAP71012
003605	000	I1=I+1	QAP71013
003606	000	INET(I1)=NCOL	QAP71014
003607	000	I=I+2	QAP71015
003608	000	IF(NCOL-NCOL1)220,700,220	QAP71016
003609	000	640 I=I-2	QAP71017
003610	000	IF(I)360,360,680	QAP71018
003611	000	680 NROW=INET(I)	QAP71019
003612	000	I1=I+1	QAP71020
003613	000	NCOL=INET(I1)	QAP71021
003614	000	INET1(NROW)=0	QAP71022
003615	000	GO TO 260	QAP71023
003616	000	C	QAP71024

003617	000	700 I=3	QAP71025
003618	000	C *) CALCULATE THE MINIMUM ENTRY IN BASIS LOOP	QAP71026
003619	000	ISAV=K2	QAP71027
003620	000	720 NROW=INET(I)	QAP71028
003621	000	I1=I+1	QAP71029
003622	000	NCOL=INET(I1)	QAP71030
003623	000	IF (JBAS(NROW,NCOL) - ISAV) 740,760,780	QAP71031
003624	000	740 ISAV=JBAS(NROW,NCOL)	QAP71032
003625	000	NROW2=NROW	QAP71033
003626	000	NCOL2=NCOL	QAP71034
003627	000	780 IF (NCOL-NCOL1) 800,860,880	QAP71035
003628	000	800 I=I+4	QAP71036
003629	000	IF (I-K) 720,720,800	QAP71037
003630	000	C *) THE BASIS LOOP HAS MORE ELEMENTS THAN THERE ARE IN THE BASIS ,	QAP71038
003631	000	C ERROR TYPE 3	QAP71039
003632	000	840 I=3	QAP71040
003633	000	C	QAP71041
003634	000	WRITE(6,3) I, NROW1, NCOL1	QAP71042
003635	000	KEY=2	QAP71043
003636	000	C	QAP71044
003637	000	RETURN	QAP71045
003638	000	C <<+++++*****	QAP71046
003639	000	860 IF (ISAV-K2) 900,880,840	QAP71047
003640	000	C *) THERE IS NO ELEMENT IN THE BASIS LOOP .LT. K. ERROR TYPE 4	QAP71048
003641	000	880 I=4	QAP71049
003642	000	C	QAP71050
003643	000	WRITE(6,4) I, NROW1, NCOL1	QAP71051
003644	000	KEY=2	QAP71052
003645	000	C	QAP71053
003646	000	RETURN	QAP71054
003647	000	C <<+++++*****	QAP71055
003648	000	C C) NOW THE LOCATIONS OF ALL ELEMENTS IN THE BASIS ARE SEQUENTIALLY	QAP71056
003649	000	C ARRANGED (PAIRWISE) IN ORDER OF I AND I1 AND ISAV CONTAINS THE MINIMUM	QAP71057
003650	000	C ENTRY IN THE LOOP , SO WE CAN CALCULATE THE BASIC SOLUTION BY	QAP71058
003651	000	C SUCCESSIVE ADDITIONS AND SUBTRACTIONS FROM THE ELEMENTS OF THE	QAP71059
003652	000	C BASIS LOOP	QAP71060
003653	000	C -----	QAP71061
003654	000	900 J=-1	QAP71062
003655	000	NROW=INET(1)	QAP71063
003656	000	NCOL=INET(2)	QAP71064
003657	000	JBAS(NROW,NCOL)=ISAV	QAP71065
003658	000	ISAV=ISAV-K1	QAP71066
003659	000	C	QAP71067
003660	000	I=3	QAP71068
003661	000	920 NROW=INET(I)	QAP71069
003662	000	I1=I+1	QAP71070
003663	000	NCOL=INET(I1)	QAP71071
003664	000	IF (NROW-NROW2) 1000,980,1000	QAP71072
003665	000	980 IF (NCOL-NCOL2) 1000,980,1000	QAP71073
003666	000	940 JBAS(NROW,NCOL)=0	QAP71074
003667	000	GO TO 1020	QAP71075
003668	000	C	QAP71076
003669	000	1000 ISET=JBAS(NROW,NCOL)+J*ISAV	QAP71077
003670	000	JBAS(NROW,NCOL)=ISET	QAP71078
003671	000	C	QAP71079
003672	000	1020 J=-J	QAP71080
003673	000	I=I+2	QAP71081



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003674      000      IF(NCOL-NCOL1)920,1000,000 QAP71082
003675      000 C <<+++++***** QAP71083
003676      000 C NOW THAT THE NEW BASIC SOLUTION HAS BEEN OBTAINED , SET KEY = 1 TO QAP71084
003677      000 C ENABLE HADY TO CHECK OPTIMALITY QAP71085
003678      000 C ----- QAP71086
003679      000     1000 KEY=1 QAP71087
003680      000     RETURN QAP71088
003681      000 C QAP71089
003682      000     1 FORMAT(10X,'ERROR TYPE 1'/10X,'NO BASIS ELEMENT IN ROW WHICH CONTAINS' QAP71090
003683      000       *INS ENTERING ELEMENT IN AN ITERATION ',3I3) QAP71091
003684      000     2 FORMAT(10X,'ERROR TYPE 2'/10X,'THERE IS NO BASIS LOOP',3I3) QAP71092
003685      000     3 FORMAT(10X,'ERROR TYPE 3'/10X,'THE BASIS LOOP HAS MORE ELEMENTS THAN' QAP71093
003686      000       *AN THERE ARE IN THE BASIS ',3I3) QAP71094
003687      000     4 FORMAT(10X,'ERROR TYPE 4'/10X,'THERE IS NO ELEMENT IN THE BASIS ;' QAP71095
003688      000       *LOOP LESS THEN K2',3I3) QAP71096
003689      000 C QAP71097
003690      000 C <<+++++***** QAP71098
003691      000     END QAP71099
003692      000 DELT,SI TUMER-S*SINAN-QAP7.1099,717053012
003693      000 C ***** QAP7 718
003694      000 C A SUBROUTINE TO FIND AN INITIAL SOLUTION OF A TRANSPORTATION PROBLEM QAP7 719
003695      000 C USING THE METHOD OF ROW PIVOTS QAP7 720
003696      000 C APPLICABLE TO DEGENERATE CASES BUT NOT TO UNDEGRADED ONES QAP7 721
003697      000 C ALL X'S HAVE TO BE UPDATED BY COMPACTING K1 QAP7 722
003698      000 C <<+++++***** QAP7 723
003699      000 C QAP7 724
003700      000     SUBROUTINE HADY(NORIG,K1,NROW1,NCOL1,ITOT,KEY) QAP7 725
003701      000 C ----- QAP7 726
003702      000     COMMON IC1,IC2,IC3,NFCB1,NFEST,NFIXD,LRCST,KTOP QAP7 727
003703      000     COMMON MSTER(06200),LRCST(0600),KSNOW(80) QAP7 728
003704      000     COMMON KLRND,KHADY,KCST,VLCST,QAP7 729
003705      000 C QAP7 730
003706      000     INTEGER*2 ICOST(40,40),TCOST(40),IDEST(40),JBNAS(40,40) QAP7 731
003707      000     DIMENSION IU(40),IV(40),IS(40),IV1(40) QAP7 732
003708      000 C QAP7 733
003709      000 C QAP7 734
003710      000     EQUIVALENCE( IWORK(0001),ICOST(5,1)) QAP7 735
003711      000     EQUIVALENCE( IWORK(0001),IU(001)) QAP7 736
003712      000     EQUIVALENCE( IWORK(0001),IDEST(001)) QAP7 737
003713      000     EQUIVALENCE( IWORK(0001), IV(001)) QAP7 738
003714      000     EQUIVALENCE( IWORK(0001), IV1(001)) QAP7 739
003715      000     EQUIVALENCE( IWORK(0001), IV1(001)) QAP7 740
003716      000     EQUIVALENCE( IWORK(0001), IV1(001)) QAP7 741
003717      000     EQUIVALENCE( IWORK(1601), JBNAS(5,1)) QAP7 742
003718      000 C <<+++++***** QAP7 743
003719      000 C UPDATE CALLS COUNTER QAP7 744
003720      000     KHADY=KHADY+1 QAP7 745
003721      000 C QAP7 746
003722      000     GO TO(320,5),KEY QAP7 747
003723      000 C ----- QAP7 748
003724      000 C INITIALIZE QAP7 749
003725      000     5 K1=30000 QAP7 750
003726      000     NDST1=NDEST QAP7 751
003727      000     NORG1=NORIG QAP7 752
003728      000     NCOL=1 QAP7 753
003729      000     NROW=1 QAP7 754
003730      000     ISAV=K1 QAP7 755

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003731	000	DO 130 I=1,NORIG	GAP7 756
003732	000	DO 130 J=1,NDEST	GAP7 757
003733	000	JBAS(I,J)=0	GAP7 758
003734	000	DO 140 J=1,NDEST	GAP7 759
003735	000	140 IV(J)=1	GAP7 760
003736	000	C FROM HERE UNTIL JUST BEFORE THE NEXT VARIABLES ARE ALLOCATED	GAP7 761
003737	000	150 IF(ICOST(NROW,NCOL)-ISAV)160,160,160	GAP7 762
003738	000	160 NCOL=NCOL+1	GAP7 763
003739	000	IF(NCOL-NDEST)150,150,END	GAP7 764
003740	000	160 IF(IV(NCOL)-1)160,160,160	GAP7 765
003741	000	190 ISAV=ICOST(NROW,NCOL)	GAP7 766
003742	000	NCOL1=NCOL	GAP7 767
003743	000	GO TO 160	GAP7 768
003744	000	C	GAP7 769
003745	000	200 IF(IORIG(NROW)-IDEST(NCOL1))210,230,310	GAP7 770
003746	000	210 ISET=IORIG(NROW)+K1	GAP7 771
003747	000	C	GAP7 772
003748	000	JBAS(NROW,NCOL1)=ISET	GAP7 773
003749	000	C	GAP7 774
003750	000	IDEST(NCOL1)=IDEST(NCOL1-10-10(NROW))	GAP7 775
003751	000	GO TO 290	GAP7 776
003752	000	C	GAP7 777
003753	000	230 ISET=IORIG(NROW)+K1	GAP7 778
003754	000	C	GAP7 779
003755	000	JBAS(NROW,NCOL1)=ISET	GAP7 780
003756	000	C	GAP7 781
003757	000	IV(NCOL1)=0	GAP7 782
003758	000	C MAKE NECESSARY ARRANGEMENTS FOR DEGENERACY	GAP7 783
003759	000	NCOL=1	GAP7 784
003760	000	ISAV=K1	GAP7 785
003761	000	IF(NROW-NORIG)240,320,END	GAP7 786
003762	000	240 IF(ICOST(NROW,NCOL)-ISAV)250,270,270	GAP7 787
003763	000	250 IF(IV(NCOL)-1)270,270,270	GAP7 788
003764	000	260 ISAV=ICOST(NROW,NCOL)	GAP7 789
003765	000	NCOL1=NCOL	GAP7 790
003766	000	270 NCOL=NCOL+1	GAP7 791
003767	000	IF(NCOL-NDEST)240,240,END	GAP7 792
003768	000	C	GAP7 793
003769	000	260 JBAS(NROW,NCOL1)=K1	GAP7 794
003770	000	C	GAP7 795
003771	000	290 NROW=NROW+1	GAP7 796
003772	000	IF(NROW-NORIG)300,300,320	GAP7 797
003773	000	300 NCOL=1	GAP7 798
003774	000	ISAV=K1	GAP7 799
003775	000	GO TO 150	GAP7 800
003776	000	C	GAP7 801
003777	000	310 ISET=IDEST(NCOL1)+K1	GAP7 802
003778	000	C	GAP7 803
003779	000	JBAS(NROW,NCOL1)=ISET	GAP7 804
003780	000	C	GAP7 805
003781	000	IORIG(NROW)=IORIG(NROW)-IDEST(NCOL1)	GAP7 806
003782	000	IV(NCOL1)=0	GAP7 807
003783	000	GO TO 300	GAP7 808
003784	000	C <<++++++>>	GAP7 809
003785	000	C START CALCULATING THE DUAL VARIABLES	GAP7 810
003786	000	C -----	GAP7 811
003787	000	C A) INITIALIZE AS FOLLOWS	GAP7 812

Address	Instruction	Hex	Dec	Comment
003780	C	* SET ALL DUAL VARIABLES IU + IV EQUAL TO ZERO	813	
003789	C	* FLAG ALL DUAL VARIABLES, IU+ IV ARE SET EQUAL TO ZERO	814	
003790		320 DO 330 I=1,NORIG	815	
003791		IU(I)=0	816	
003792		330 IU1(I)=0	817	
003793		DO 340 J=1,NDEST	818	
003794		IV(J)=0	819	
003795		340 IV1(J)=0	820	
003796	C	UNFLAG THE FIRST ROW	821	
003797		IU1(1)=1	822	
003798		NROW=1	823	
003799		NCOL=1	824	
003800	C	B) FROM HERE UNTIL JUST BEFORE 540 THE REMAINING DUAL VARIABLES ARE	825	
003801	C	DETERMINED BY REFILLING SCANNING	826	
003802		360 IF(JBAS(NROW,NCOL)-1)370,370,370	827	
003803		370 IV(NCOL)=ICOST(NROW,NCOL)-IU(NROW)	828	
003804	C	* UNFLAG COLUMN NCOL	829	
003805		IV1(NCOL)=1	830	
003806		380 NCOL=NCOL+1	831	
003807		IF(NCOL-NDEST)360,360,400	832	
003808		400 NROW=NROW+1	833	
003809		IF(NROW-NORIG)420,420,500	834	
003810		420 IF(IU1(NROW)-1)430,400,400	835	
003811		430 NCOL=1	836	
003812		440 IF(JBAS(NROW,NCOL)-K1)450,470,470	837	
003813		450 NCOL=NCOL+1	838	
003814		IF(NCOL-NDEST)440,400,400	839	
003815		470 IF(IV1(NCOL))480,450,450	840	
003816		480 IU(NROW)=ICOST(NROW,NCOL)-IV(NCOL)	841	
003817	C	* UNFLAG ROW NROW	842	
003818		IU1(NROW)=1	843	
003819		NCOL=1	844	
003820		GO TO 360	845	
003821	C	* CHECK WHETHER ANY ROWS ARE STILL FLAGGED , IF SO RETURN TO 430 ,	846	
003822	C	OTHERWISE GO TO 540	847	
003823		500 NROW=1	848	
003824		510 IF(IU1(NROW))520,430,500	849	
003825		520 NROW=NROW+1	850	
003826		IF(NROW-NORIG)510,510,500	851	
003827	C	<<+++++>>	852	
003828	C	CHECK OPTIMALITY OF THE CURRENT SOLUTION	853	
003829	C	-----	854	
003830		540 ISAV=0	855	
003831		NROW=1	856	
003832		550 NCOL=1	857	
003833		560 IF(JBAS(NROW,NCOL)-K1)570,570,570	858	
003834		570 NCOL=NCOL+1	859	
003835		IF(NCOL-NDEST)560,560,500	860	
003836		590 NROW=NROW+1	861	
003837		IF(NROW-NORIG)550,550,610	862	
003838		610 IF(ISAV)700,800,700	863	
003839		620 IS1=IU(NROW)+IV(NCOL)-ICOST(NROW,NCOL)	864	
003840		IF(IS1-ISAV)570,570,640	865	
003841		640 ISAV=IS1	866	
003842	C		867	
003843		NROW1=NROW	868	
003844		NCOL1=NCOL	869	

003845	000		GO TO 370	GAP7 870
003846	000	C	<<+++++>>	GAP7 871
003847	000	C	CURRENT SOLUTION IS NORMAL	GAP7 872
003848	000	C	-----	GAP7 873
003849	000		700 KEY=1	GAP7 874
003850	000		RETURN	GAP7 875
003851	000	C	<<+++++>>	GAP7 876
003852	000	C	CURRENT SOLUTION IS OPTIMAL	GAP7 877
003853	000	C	-----	GAP7 878
003854	000		800 KEY=2	GAP7 879
003855	000	C	CALCULATE ACTUAL VALUES OF BASIC VARIABLES	GAP7 880
003856	000		ITOT=0	GAP7 881
003857	000		DO 850 I=1,NORIG	GAP7 882
003858	000		DO 830 J=1,NDELT	GAP7 883
003859	000		IF(JBAS(I,J))850,850,850	GAP7 884
003860	000	820	JBASE =JBAS(I,J)+1	GAP7 885
003861	000		ITOT=ITOT+(JBASE - 100*(I+J))	GAP7 886
003862	000	830	CONTINUE	GAP7 887
003863	000	850	CONTINUE	GAP7 888
003864	000		IF(KNSNW(30).EQ.0)GO TO 900	GAP7 889
003865	000		WRITE(6,11)	GAP7 890
003866	000		DO 870 I=1,NORIG	GAP7 891
003867	000	870	WRITE(6,01)I,(JBAS(I,J),J+1,NDELT)	GAP7 892
003868	000	C		GAP7 893
003869	000		900 RETURN	GAP7 894
003870	000	C		GAP7 895
003871	000		1 FORMAT(10X,I3,2X,4011)	GAP7 896
003872	000	11	FORMAT(/10X,'OPTIMAL SOLUTION OF THE LINEAR ASSIGNMENT PROBLEM')	GAP7 897
003873	000		END	
003874	000		MELT,S1 TUMER-S*SINAH-GAPTLENTA-106247605E012	
003875	000	C	A SUBROUTINE TO CALCULATE THE ORDER BOUND AT NEW NODE	GAP7 405
003876	000		SUBROUTINE LMBND(LBND,KHADV,KHADY,MOVE,KAT)	GAP7 406
003877	000		COMMON IC1,IC2,IC3,IC4,IC5,IC6,IC7,IC8,IC9,IC10,IC11,IC12	GAP7 407
003878	000		COMMON MSTP(106200),F(40,40),IDEST(40)	GAP7 408
003879	000		COMMON KLKND,KHADY,KHADV,FLGINT	GAP7 409
003880	000	C		GAP7 410
003881	000		INTEGER*2 WORK1(40,40),WORK2(40,40),WORK3(40),WORK4(40),WORK6(40)	GAP7 411
003882	000		INTEGER*2 WORK8(40)	GAP7 412
003883	000		INTEGER*2 F(40,40),IDEST(40),INTR1(40,40),INTR2(40,40),INTR3(40,40)	GAP7 413
003884	000		INTEGER*2 INTVL(40,40),IDEST(40),JBAS(40,40),IDEST(40)	GAP7 414
003885	000		LOGICAL*1 ASIGN(40,40),FLGINT(40,40),FLGINT(40),FLGST(40)	GAP7 415
003886	000		LOGICAL*1 FCFLG(40),BAIR(100,40)	GAP7 416
003887	000		EQUIVALENCE( IWORK(0001),WORK1(1,1))	GAP7 417
003888	000		EQUIVALENCE( IWORK(0001),WORK2(1,1))	GAP7 418
003889	000		EQUIVALENCE( IWORK(0001),WORK3(001))	GAP7 419
003890	000		EQUIVALENCE( IWORK(0001),WORK4(001))	GAP7 420
003891	000		EQUIVALENCE( IWORK(1001),WORK5(1,1))	GAP7 421
003892	000		EQUIVALENCE( IWORK(1001),WORK6( 1))	GAP7 422
003893	000		EQUIVALENCE( IWORK(1001),WORK8( 1))	GAP7 423
003894	000		EQUIVALENCE( IWORK(1001),WORK8( 1))	GAP7 424
003895	000		EQUIVALENCE( IWORK(1001),WORK8( 1))	GAP7 425
003896	000	C		GAP7 426
003897	000		EQUIVALENCE( MSTP(0001),F(1,1))	GAP7 427
003898	000		EQUIVALENCE( MSTP(0001),IDEST(1,1))	GAP7 428
003899	000		EQUIVALENCE( MSTP(1001),IDEST(1,1))	GAP7 429
003900	000		EQUIVALENCE( MSTP(2001),IDEST(1,1))	GAP7 430
003901	000		EQUIVALENCE( MSTP(3001),IDEST(1,1))	GAP7 431

003902	000	EQUIVALENCE( MSTER(6001) )=1000(11)	GAP7 432
003903	000	EQUIVALENCE( MSTER(6001) )=1000(11)	GAP7 433
003904	000	EQUIVALENCE( MSTER(6001) )=1000(11)	GAP7 434
003905	000	EQUIVALENCE( MSTER(6001) )=1000(11)	GAP7 435
003906	000	EQUIVALENCE( MSTER(6001) )=1000(11)	GAP7 436
003907	000	EQUIVALENCE( MSTER(6001) )=1000(11)	GAP7 437
003908	000	EQUIVALENCE( MSTER(6001) )=1000(11)	GAP7 438
003909	000	C	GAP7 439
003910	000	C IPDATE CALLS COUNTERS	GAP7 440
003911	000	KLBND=KLBND+1	GAP7 441
003912	000	C	GAP7 442
003913	000	IF(KSNSW(21).EQ.1.AND.MOVE.EQ.0)WRITE(3,28)	GAP7 443
003914	000	IF(KSNSW(21).EQ.1.AND.MOVE.EQ.1)WRITE(3,29)ITER,LEVY	GAP7 444
003915	000	IF(KSNSW(21).EQ.1.AND.MOVE.EQ.2)WRITE(3,31)ITER,LEVY	GAP7 445
003916	000	C	GAP7 446
003917	000	C	GAP7 447
003918	000	GO TO (130,222),KAM	GAP7 448
003919	000	130 NMIN1=NFCIL-1	GAP7 449
003920	000	NMIN2=NLOCT-1	GAP7 450
003921	000	LBTC=LBCST	GAP7 451
003922	000	KEY=2	GAP7 452
003923	000	C	GAP7 453
003924	000	C	GAP7 454
003925	000	DO 190 I=1,NFCIL	GAP7 455
003926	000	DO 165 K=1,NFCIL	GAP7 456
003927	000	WORK3(K)=F(I,K)	GAP7 457
003928	000	WORK4(K)=K	GAP7 458
003929	000	165 CONTINUE	GAP7 459
003930	000	DO 180 J=1,NMIN1	GAP7 460
003931	000	K=J+1	GAP7 461
003932	000	DO 170 L=K,NFCIL	GAP7 462
003933	000	IF(WORK3(J).GT.WORK3(L))GO TO 170	GAP7 463
003934	000	KEEP=WORK3(J)	GAP7 464
003935	000	WORK3(J)=WORK3(L)	GAP7 465
003936	000	WORK3(L)=KEEP	GAP7 466
003937	000	KEEP=WORK4(J)	GAP7 467
003938	000	WORK4(J)=WORK4(L)	GAP7 468
003939	000	WORK4(L)=KEEP	GAP7 469
003940	000	170 CONTINUE	GAP7 470
003941	000	180 CONTINUE	GAP7 471
003942	000	DO 185 J=1,NFCIL	GAP7 472
003943	000	185 INTRI(I,J)=WORK4(J)	GAP7 473
003944	000	IF(KSNSW(23).EQ.1)WRITE(6,11)I,(WORK3(L),L=1,NFCIL)	GAP7 474
003945	000	190 CONTINUE	GAP7 475
003946	000	C	GAP7 476
003947	000	C	GAP7 477
003948	000	C	GAP7 478
003949	000	DO 230 JJ=1,NLOCT	GAP7 479
003950	000	DO 192 K=1,NLOCT	GAP7 480
003951	000	WORK6(K)=D(JJ,K)	GAP7 481
003952	000	IF(WORK6(K).EQ.0)WORK6(K)=9999	GAP7 482
003953	000	WORK4(K)=K	GAP7 483
003954	000	192 CONTINUE	GAP7 484
003955	000	C	GAP7 485
003956	000	DO 210 J=1,NMIN2	GAP7 486
003957	000	K=J+1	GAP7 487
003958	000	DO 200 L=K,NLOCT	GAP7 488

003959	000	IF(WORK6(J).LT.WORK6(L))GO TO 260	QAP7 489
003960	000	KEEP=WORK6(J)	QAP7 490
003961	000	WORK6(J)=WORK6(L)	QAP7 491
003962	000	WORK6(L)=KEEP	QAP7 492
003963	000	KEEP=WORK4(J)	QAP7 493
003964	000	WORK4(J)=WORK4(L)	QAP7 494
003965	000	WORK4(L)=KEEP	QAP7 495
003966	000	200 CONTINUE	QAP7 496
003967	000	210 CONTINUE	QAP7 497
003968	000	IF(KSN5W(23).EQ.1)WRITE(6,12)JJ,(WORK6(L),L=1,NLOCT)	QAP7 498
003969	000	DO 220 K=1,NLOCT	QAP7 499
003970	000	WORK2(JJ,K)=WORK6(K)	QAP7 500
003971	000	220 DISTI(JJ,K)=WORK4(K)	QAP7 501
003972	000	230 CONTINUE	QAP7 502
003973	000	C	QAP7 503
003974	000	C	QAP7 504
003975	000	C CONSTRUCT THE WORK1 MATRIX	QAP7 505
003976	000	222 DO 320 I=1,NFCIL	QAP7 506
003977	000	IF(FCFLG(I))GO TO 300	QAP7 507
003978	000	KEEP=0	QAP7 508
003979	000	DO 250 L=1,NFCIL	QAP7 509
003980	000	LL=INTRI(I,L)	QAP7 510
003981	000	IF(FCFLG(LL))GO TO 250	QAP7 511
003982	000	KEEP=KEEP+1	QAP7 512
003983	000	WORK3(KEEP)=F(I,LL)	QAP7 513
003984	000	250 CONTINUE	QAP7 514
003985	000	NFREE=KEEP	QAP7 515
003986	000	C	QAP7 516
003987	000	DO 290 J=1,NLOCT	QAP7 517
003988	000	IF(BANED(I,J))GO TO 260	QAP7 518
003989	000	IF(FLOGT(J))GO TO 260	QAP7 519
003990	000	KEEP=0	QAP7 520
003991	000	DO 260 L=1,NLOCT	QAP7 521
003992	000	LL=DISTI(J,L)	QAP7 522
003993	000	IF(FLOGT(LL))GO TO 260	QAP7 523
003994	000	KEEP=KEEP+1	QAP7 524
003995	000	WORK4(KEEP)=D(J,LL)	QAP7 525
003996	000	260 CONTINUE	QAP7 526
003997	000	KEEP=0	QAP7 527
003998	000	IF(NFREE.EQ.0)GO TO 272	QAP7 528
003999	000	C HERE WE CALCULATE WHAT IS EQUIVALENT TO AN OLD 103 ELEMENT	QAP7 529
004000	000	DO 270 L=1,NFREE	QAP7 530
004001	000	270 KEEP=KEEP+(WORK3(L)+WORK4(L))	QAP7 531
004002	000	C HERE WE CALCULATE WHAT IS EQUIVALENT TO AN OLD 102 MATRIX ENTRY	QAP7 532
004003	000	272 LL=0	QAP7 533
004004	000	DO 275 L=1,NFCIL	QAP7 534
004005	000	IF(.NOT.FCFLG(L))GO TO 275	QAP7 535
004006	000	JL=LOCAT(L)	QAP7 536
004007	000	LL=LL+(D(JL,J)*F(L,I))	QAP7 537
004008	000	275 CONTINUE	QAP7 538
004009	000	C	QAP7 539
004010	000	WORK1(I,J)=KEEP+(2*LL)	QAP7 540
004011	000	GO TO 290	QAP7 541
004012	000	280 WORK1(I,J)=32767	QAP7 542
004013	000	290 CONTINUE	QAP7 543
004014	000	GO TO 320	QAP7 544
004015	000	C	QAP7 545

004016	000	300 DO 310 L=1,NLOCT	QAP7 546
004017	000	310 WORK1(I,L)=32767	QAP7 547
004018	000	L=LOCAT(I)	QAP7 548
004019	000	WORK1(I,L)=0	QAP7 549
004020	000	320 CONTINUE	QAP7 550
004021	000	C	QAP7 551
004022	000	C CALCULATE THE REDUCTIONS	QAP7 552
004023	000	KEEP=0	QAP7 553
004024	000	DO 350 I=1,NFCIL	QAP7 554
004025	000	WORK3(I)=32767	QAP7 555
004026	000	DO 330 J=1,NLOCT	QAP7 556
004027	000	IF(WORK1(I,J).GE.WORK3(I))GO TO 330	QAP7 557
004028	000	WORK3(I)=WORK1(I,J)	QAP7 558
004029	000	330 CONTINUE	QAP7 559
004030	000	KEEP=KEEP+WORK3(I)	QAP7 560
004031	000	DO 340 J=1,NLOCT	QAP7 561
004032	000	WORK2(I,J)=WORK1(I,J)	QAP7 562
004033	000	340 WORK1(I,J)=WORK1(I,J)-WORK3(I)	QAP7 563
004034	000	350 CONTINUE	QAP7 564
004035	000	C	QAP7 565
004036	000	DO 400 J=1,NLOCT	QAP7 566
004037	000	WORK6(J)=32767	QAP7 567
004038	000	DO 375 I=1,NFCIL	QAP7 568
004039	000	IF(WORK1(I,J).GE.WORK6(J))GO TO 375	QAP7 569
004040	000	WORK6(J)=WORK1(I,J)	QAP7 570
004041	000	375 CONTINUE	QAP7 571
004042	000	DO 380 I=1,NFCIL	QAP7 572
004043	000	380 WORK1(I,J)=WORK1(I,J)-WORK6(J)	QAP7 573
004044	000	400 CONTINUE	QAP7 574
004045	000	IF(NFCIL.EQ.NLOCT)GO TO 410	QAP7 575
004046	000	DO 405 I=1,10	QAP7 576
004047	000	405 WRITE(6,98)	QAP7 577
004048	000	98 FORMAT(02X,2(' **** WARNING ****'),3X,' YOU NEED TO ARRANGE AN	QAP7 578
004049	000	* ARRAY IN LWBND2 SO INCLUDE THE SUBROUTINE')	QAP7 579
004050	000	STOP	QAP7 580
004051	000	410 DO 415 J=1,NFCIL	QAP7 581
004052	000	415 KEEP=KEEP+WORK6(J)	QAP7 582
004053	000	LBCST=KEEP	QAP7 583
004054	000	C	QAP7 584
004055	000	C PRINT THE MATRICES IF YOU WANT	QAP7 585
004056	000	IF(KSNSW(24).EQ.0)GO TO 420	QAP7 586
004057	000	WORK4(1)=9999	QAP7 587
004058	000	DO 420 J=2,40	QAP7 588
004059	000	420 WORK4(J)=J-1	QAP7 589
004060	000	WRITE(6,16)	QAP7 590
004061	000	WRITE(6,01)(WORK4(J),J=1,40)	QAP7 591
004062	000	WRITE(6,07)	QAP7 592
004063	000	DO 430 I=1,NFCIL	QAP7 593
004064	000	430 WRITE(6,01)I,(WORK2(I,J),J=1,NLOCT),WORK3(I)	QAP7 594
004065	000	WRITE(6,07)	QAP7 595
004066	000	WRITE(6,01)NFCIL,(WORK6(J),J=1,NLOCT),KEEP	QAP7 596
004067	000	WRITE(6,07)	QAP7 597
004068	000	WRITE(6,18)	QAP7 598
004069	000	WRITE(6,01)(WORK4(J),J=1,40)	QAP7 599
004070	000	WRITE(6,07)	QAP7 600
004071	000	DO 440 I=1,NFCIL	QAP7 601
004072	000	440 WRITE(6,01)I,(WORK1(I,J),J=1,NLOCT),WORK3(I)	QAP7 602

004073	000	WRITE(6,07)	GAP7 603
004074	000	WRITE(6,01)NFCIL,(WORK6(J),J=1,NLOCT),KEEP	GAP7 604
004075	000	WRITE(6,07)	GAP7 605
004076	000	C	GAP7 606
004077	000	450 IF(KSNSW(25).EQ.1)GO TO 773	GAP7 607
004078	000	IF(KEEP.GE.32767)GO TO 773	GAP7 608
004079	000	IF(KAM.EQ.1)GO TO 773	GAP7 609
004080	000	KOMP=0.5*FLOAT(KEEP)	GAP7 610
004081	000	KOMP=KOMP+IC1+IWD	GAP7 611
004082	000	IF(KOMP.LT.KTOT)GO TO 773	GAP7 612
004083	000	ITOT=0	GAP7 613
004084	000	IF(KSNSW(28).EQ.1)WRITE(6,28)KOMP,KTOT	GAP7 614
004085	000	NECTP=NECTP+1	GAP7 615
004086	000	GO TO 783	GAP7 616
004087	000	465 LBCST=32767	GAP7 617
004088	000	IC2=32767	GAP7 618
004089	000	LTOT=3276700	GAP7 619
004090	000	GO TO 820	GAP7 620
004091	000	C	GAP7 621
004092	000	C INITIALIZE HADY	GAP7 622
004093	000	750 N1=NFCIL	GAP7 623
004094	000	IF(NFCIL.EQ.NLOCT)GO TO 773	GAP7 624
004095	000	C ADD ADUMMY FACILITY	GAP7 625
004096	000	N1=NFCIL+1	GAP7 626
004097	000	DO 755 J=1,NLOCT	GAP7 627
004098	000	755 WORK1(N1,J)=0	GAP7 628
004099	000	TOPIC(N1)=NLOCT-NFCIL	GAP7 629
004100	000	760 DO 765 I=1,NFCIL	GAP7 630
004101	000	765 TOPIC(I)=1	GAP7 631
004102	000	DO 770 J=1,NLOCT	GAP7 632
004103	000	770 IDEST(J)=1	GAP7 633
004104	000	KEY=2	GAP7 634
004105	000	773 CALL HADY(N1,K1,BROW1,WORK1,IDEST,KEY)	GAP7 635
004106	000	GO TO(775,783),KEY	GAP7 636
004107	000	775 CALL GARY(N1,K1,BROW1,WORK1,KEY)	GAP7 637
004108	000	GO TO(773,780),KEY	GAP7 638
004109	000	780 WRITE(6,99)	GAP7 639
004110	000	RETURN	GAP7 640
004111	000	99 FORMAT(02X,3(' ***** ERROR IN GARY ***'))	GAP7 641
004112	000	783 MDCST=ITOT+LBCST	GAP7 642
004113	000	NECTP=NECTP+1	GAP7 643
004114	000	IF(KSNSW(28).EQ.1)WRITE(6,28)ITOT	GAP7 644
004115	000	C	GAP7 645
004116	000	IC2=0.5*FLOAT(MDCST)	GAP7 646
004117	000	KEY=1	GAP7 647
004118	000	LTOT=0	GAP7 648
004119	000	C EVALUATE THE SOLUTION	GAP7 649
004120	000	DO 784 K=1,NLOCT	GAP7 650
004121	000	784 WORK8(K)=0	GAP7 651
004122	000	DO 790 I=1,NFCIL	GAP7 652
004123	000	DO 785 K=1,NLOCT	GAP7 653
004124	000	IF(WORK8(K).EQ.1)GO TO 790	GAP7 654
004125	000	IF(JBAS(I,K).LE.K1)GO TO 785	GAP7 655
004126	000	WORK4(I)=K	GAP7 656
004127	000	WORK8(K)=1	GAP7 657
004128	000	GO TO 790	GAP7 658
004129	000	785 CONTINUE	GAP7 659



Line	Code	Statement	Page
004130	000	790 CONTINUE	QAP7 660
004131	000	C	QAP7 661
004132	000	DO 810 I=1,NFCIL	QAP7 662
004133	000	DO 800 J=1,NFCIL	QAP7 663
004134	000	LOC1=WORK4(I)	QAP7 664
004135	000	LOCJ=WORK4(J)	QAP7 665
004136	000	800 LTOT=LTOT+(F(I,J)*D(LOC1,LOCJ))	QAP7 666
004137	000	810 CONTINUE	QAP7 667
004138	000	LTOT=0.5*FLOAT(LTOT)	QAP7 668
004139	000	IF(KAM.EQ.1)KTOT=LTOT	QAP7 669
004140	000	C	QAP7 670
004141	000	C UPDATE IC1	QAP7 671
004142	000	820 IC1=IC1+IWD	QAP7 672
004143	000	IF(KAM.EQ.2)GO TO 830	QAP7 673
004144	000	KEEP=IC1+IC2	QAP7 674
004145	000	IF(KEEP.GT.LBTC )GO TO 830	QAP7 675
004146	000	DO 825 I=1,10	QAP7 676
004147	000	825 WRITE(6,41)KEEP,LBTC	QAP7 677
004148	000	LBCST=LBTC	QAP7 678
004149	000	GO TO 840	QAP7 679
004150	000	830 LBCST=IC1+IC2	QAP7 680
004151	000	C	QAP7 681
004152	000	840 IF(KSNSW(22).EQ.1)WRITE(6,30)IC1	QAP7 682
004153	000	IF(KSNSW(22).EQ.1)WRITE(6,30)IC2	QAP7 683
004154	000	IF(KSNSW(26).EQ.1)WRITE(6,30)KTOT	QAP7 684
004155	000	IF(KSNSW(29).EQ.1)WRITE(6,30)LBCST	QAP7 685
004156	000	C	QAP7 686
004157	000	RETURN	QAP7 687
004158	000	C	QAP7 688
004159	000	1 FORMAT(10X,20I4)	QAP7 689
004160	000	6 FORMAT(1H1)	QAP7 690
004161	000	7 FORMAT(/)	QAP7 691
004162	000	11 FORMAT(/10X,'INTERACTIONS FOR ELEMENT',I3,2X,20I4)	QAP7 692
004163	000	12 FORMAT(/10X,'DISTANCES FOR LOCATION ',I3,2X,20I4)	QAP7 693
004164	000	14 FORMAT(1H+,'110X','PRODUCTS',I3)	QAP7 694
004165	000	16 FORMAT(/02X,'MATRICES FOR LOWER BOUND FOLLOW,/'	QAP7 695
004166	000	* 10X,'COMPLETE MATRIX',I3)	QAP7 696
004167	000	18 FORMAT(10X,'REDUCED MATRIX',I3)	QAP7 697
004168	000	19 FORMAT(100X,'ABSOLUTE LOWER BOUND=',I8)	QAP7 698
004169	000	21 FORMAT(/10X,'MATRIX OF LOWER BOUNDS WHEN INEQUALITIES ARE ENFORCED',I3)	QAP7 699
004170	000	22 FORMAT(/10X,'MATRIX OF LOWER BOUNDS WHEN EXCLUSIONS ARE ENFORCED',I3)	QAP7 700
004171	000	23 FORMAT(/10X,'LOWER BOUND CALCULATED BY LP ',I3)	QAP7 701
004172	000	24 FORMAT(10X,'NOT NECESSARY TO SOLVE TP AS LB BY REDUCT=',I8,' WHEN',I3)	QAP7 702
004173	000	*E CURRENT UB=',I8)	QAP7 703
004174	000	28 FORMAT(1H0,' RESULTS FOR CALCULATIONS FOR LB AT THE START')	QAP7 704
004175	000	29 FORMAT(1H0,' RESULTS FOR CALCULATIONS FOR LB AT ITER ',I3,'	QAP7 705
004176	000	*LEVEL',I3,' ON A FORWARD PASS')	QAP7 706
004177	000	31 FORMAT(1H0,' RESULTS FOR CALCULATIONS FOR LB AT ITER ',I3,'	QAP7 707
004178	000	*LEVEL',I3,' ON A BACKWARD PASS')	QAP7 708
004179	000	32 FORMAT(90X,'VALUE IF C1 =',I3)	QAP7 709
004180	000	34 FORMAT(90X,'VALUE OF C2 =',I3)	QAP7 710
004181	000	36 FORMAT(90X,'VALUE OF UB =',I3,' I7)	QAP7 711
004182	000	39 FORMAT(90X,'VALUE OF LB =',I3)	QAP7 712
004183	000	41 FORMAT(02X,4('**** NOOOOOOUTICE ****'),' IC1+IC2=',I7,2X,'LBCST	QAP7 713
004184	000	*READ IN=',I7)	QAP7 714
004185	000	C	QAP7 715
004186	000	C	QAP7 716

Line	Code	Statement	Gain
004187	000	END	
004188	000	DEL,SI TUMER-S*SINAN-QAP7,TIMEOUT,,,152574103711	QAP7 717
004189	000	SUBROUTINE TIMEOUT	
004190	000	K=ITIME(I,J)	
004191	000	T=1/5000.	
004192	000	WRITE(6,2000)T	
004193	000	2000 FORMAT(' TIME',F10.3,' SEC')	
004194	000	RETURN	
004195	000	END	
004196	000		
004197	000		
004198	000	DEL,SI TUMER-S*SINAN-QAIN-QAP7,121364061212	
004199	000	COMMON IC1,IC2,IC3,IC4,IC5,IC6,IC7,MAXM,LLCCT	GAIN 1
004200	000	COMMON MSTP(03400),IPR,PC(120),KSNWSW(80)	GAIN 2
004201	000	INTEGER F(40,40),F1(40,40)	GAIN 3
004202	000	INTEGER FIX(40),F2(40)	GAIN 4
004203	000	INTEGER WORK1(40),WORK2(40),WORK3(40),F3(40),F4(40)	GAIN 5
004204	000	LOGICAL FCFLG(40)	GAIN 6
004205	000	INTEGER WORK4(40),WORK5(40),WORK6(40),WORK7(40)	GAIN 7
004206	000	INTEGER IFLAG(40),WORK8(40),WORK9(40),WORK10(40)	GAIN 8
004207	000	INTEGER WORK11(40),WORK12(40),WORK13(40),WORK14(40)	GAIN 9
004208	000	INTEGER CSTM(40),CSTN(40),CSTP(40),CSTQ(40)	
004209	000	C EQUIVALENCE( MSTP(40),F(40,40))	GAIN 10
004210	000	EQUIVALENCE( MSTP(40),F1(40,40))	GAIN 11
004211	000	EQUIVALENCE( MSTP(40),F2(40))	GAIN 12
004212	000	EQUIVALENCE( MSTP(40),F3(40))	GAIN 13
004213	000	EQUIVALENCE( MSTP(40),F4(40))	GAIN 14
004214	000	EQUIVALENCE( MSTP(40),F5(40))	GAIN 15
004215	000	EQUIVALENCE( MSTP(40),F6(40))	GAIN 16
004216	000	EQUIVALENCE( MSTP(40),F7(40))	GAIN 17
004217	000	C EQUIVALENCE( IWORK(40),F(40,40))	GAIN 18
004218	000	EQUIVALENCE( IWORK(40),F1(40,40))	GAIN 19
004219	000	EQUIVALENCE( IWORK(40),F2(40))	GAIN 20
004220	000	EQUIVALENCE( IWORK(40),F3(40))	GAIN 21
004221	000	EQUIVALENCE( IWORK(40),F4(40))	GAIN 22
004222	000	EQUIVALENCE( IWORK(40),F5(40))	GAIN 23
004223	000	EQUIVALENCE( IWORK(40),F6(40))	GAIN 24
004224	000	EQUIVALENCE( IWORK(40),F7(40))	GAIN 25
004225	000	EQUIVALENCE( IWORK(40),F8(40))	GAIN 26
004226	000	EQUIVALENCE( IWORK(40),F9(40))	GAIN 27
004227	000	EQUIVALENCE( IWORK(40),F10(40))	GAIN 28
004228	000	EQUIVALENCE( IWORK(40),F11(40))	GAIN 29
004229	000	EQUIVALENCE( IWORK(40),F12(40))	GAIN 30
004230	000	EQUIVALENCE( IWORK(40),F13(40))	GAIN 31
004231	000	EQUIVALENCE( IWORK(40),F14(40))	GAIN 32
004232	000	C 10 CALL DTAIN(JOB,IFRAC,IPR,PC,ICOTA)	GAIN 33
004233	000	C	GAIN 34
004234	000	C	GAIN 35
004235	000	C	GAIN 36
004236	000	C PRINT THE INITIAL DATA IF YOU WISH	GAIN 37
004237	000	WRITE(6,4)JOB,IFRAC,IPR,PC	GAIN 38
004238	000	IF(KSNWSW(01).EQ.0)GO TO 110	GAIN 39
004239	000	DO 70 I=1,02	GAIN 40
004240	000	70 WRITE(6,2)	GAIN 41
004241	000	WRITE(6,5)	GAIN 42
004242	000	DO 80 I=1,NLOCT	GAIN 43
004243	000	80 WRITE(6,01)(D(I,J),J=1,NLOCT)	GAIN 44

000244	000	WRITE(6,8)	GAIN	45
000245	000	DO 90 I=1,NFCIL	GAIN	46
000246	000	90 WRITE(6,01)(F(I,J),J=1,NFCIL)	GAIN	47
000247	000	C	GAIN	48
000248	000	110 NMIN1=NLOCT-1	GAIN	49
000249	000	NMIN2=NFCIL-1	GAIN	50
000250	000	C	GAIN	51
000251	000	DO 115 I=1,40	GAIN	52
000252	000	LOCAT(I)=32767	GAIN	53
000253	000	DO 115 J=1,40	GAIN	54
000254	000	115 MAKAN(I,J)=32767	GAIN	55
000255	000	C CALCULATE THE DISTANCE FROM EACH LOCATION TO ALL OTHER LOCATIONS	GAIN	56
000256	000	DO 130 I=1,NLOCT	GAIN	57
000257	000	WORK2(I)=0	GAIN	58
000258	000	WORK3(I)=I	GAIN	59
000259	000	DO 130 J=1,NLOCT	GAIN	60
000260	000	130 WORK2(I)=WORK2(I)+D(I,J)	GAIN	61
000261	000	IF(KSNSW(21).EQ.1)GO TO 140,11(WORK2(I),I=NLOCT)	GAIN	62
000262	000	IF(KSNSW(21).EQ.1)GO TO 140,11(WORK3(I),I=NLOCT)	GAIN	63
000263	000	C RANK ALL THE LOCATIONS IN AN ASCENDING ORDER OF THE DISTANCES	GAIN	64
000264	000	DO 180 I=1,NMIN1	GAIN	65
000265	000	K=I+1	GAIN	66
000266	000	DO 170 J=K,NLOCT	GAIN	67
000267	000	IF(WORK2(I).LT.WORK2(J))GO TO 170	GAIN	68
000268	000	KEEP=WORK2(I)	GAIN	69
000269	000	LEEP=WORK3(I)	GAIN	70
000270	000	WORK2(I)=WORK2(J)	GAIN	71
000271	000	WORK3(I)=WORK3(J)	GAIN	72
000272	000	WORK2(J)=KEEP	GAIN	73
000273	000	WORK3(J)=LEEP	GAIN	74
000274	000	170 CONTINUE	GAIN	75
000275	000	180 CONTINUE	GAIN	76
000276	000	DO 190 J=1,NLOCT	GAIN	77
000277	000	K=WORK3(J)	GAIN	78
000278	000	JFLAG(K)=J	GAIN	79
000279	000	190 CONTINUE	GAIN	80
000280	000	C	GAIN	81
000281	000	C	GAIN	82
000282	000	IF(KSNSW(22).EQ.0)GO TO 200	GAIN	83
000283	000	WRITE(6,21)	GAIN	84
000284	000	WRITE(6,11)(WORK2(I),I=1,NLOCT)	GAIN	85
000285	000	WRITE(6,12)(WORK3(I),I=1,NLOCT)	GAIN	86
000286	000	WRITE(6,07)	GAIN	87
000287	000	WRITE(6,14)(JFLAG(I),I=1,NLOCT)	GAIN	88
000288	000	C	GAIN	89
000289	000	C START RANKING THE FACILITIES	GAIN	90
000290	000	C A) CALCULATE THE NUMBER OF ELEMENTS AND CONNECTIONS WITH EACH ELEMENT	GAIN	91
000291	000	200 DO 320 I=1,NFCIL	GAIN	92
000292	000	WORK7(I)=I	GAIN	93
000293	000	WORK4(I)=I	GAIN	94
000294	000	WORK5(I)=I	GAIN	95
000295	000	E1(I)=0	GAIN	96
000296	000	E2(I)=0	GAIN	97
000297	000	DO 310 J=1,NFCIL	GAIN	98
000298	000	IF(I.EQ.J)GO TO 310	GAIN	99
000299	000	ISAV=F(I,J)	GAIN	100
000300	000	IF(ISAV.LT.F(J,I))ISAV=F(J,I)	GAIN	101

004301	000	IF(ISAV.EQ.0)GO TO 310	GAIN 102
004302	000	E1(I)=E1(I)+1	GAIN 103
004303	000	E2(I)=E2(I)+ISAV	GAIN 104
004304	000	310 CONTINUE	GAIN 105
004305	000	320 CONTINUE	GAIN 106
004306	000	C	GAIN 107
004307	000	C	GAIN 108
004308	000	IF(KSNSW(23).EQ.0)GO TO 400	GAIN 109
004309	000	WRITE(6,12)(WORK4(I),I=1,NFCIL)	GAIN 110
004310	000	WRITE(6,17)( E1(I),I=1,NFCIL)	GAIN 111
004311	000	WRITE(6,18)( E2(I),I=1,NFCIL)	GAIN 112
004312	000	C	GAIN 113
004313	000	C RANK THE ELEMENTS IN A DESCENDING ORDER OF THE NUMBER OF ELEMENTS	GAIN 114
004314	000	C CONNECTED TO EACH	GAIN 115
004315	000	330 DO 360 I=1,NMIN2	GAIN 116
004316	000	K=I+1	GAIN 117
004317	000	DO 350 J=K,NFCIL	GAIN 118
004318	000	IF(E1(I).GT.E1(J))GO TO 350	GAIN 119
004319	000	KEEP=E1(I)	GAIN 120
004320	000	LEEP=WORK4(I)	GAIN 121
004321	000	E1(I)=E1(J)	GAIN 122
004322	000	E1(J)=KEEP	GAIN 123
004323	000	WORK4(I)=WORK4(J)	GAIN 124
004324	000	WORK4(J)=LEEP	GAIN 125
004325	000	350 CONTINUE	GAIN 126
004326	000	360 CONTINUE	GAIN 127
004327	000	C	GAIN 128
004328	000	C	GAIN 129
004329	000	IF(KSNSW(24).EQ.0)GO TO 400	GAIN 130
004330	000	WRITE(6,26)	GAIN 131
004331	000	WRITE(6,17)( E1(I),I=1,NFCIL)	GAIN 132
004332	000	WRITE(6,12)(WORK4(I),I=1,NFCIL)	GAIN 133
004333	000	C	GAIN 134
004334	000	C RANK THE ELEMENTS IN A DESCENDING ORDER OF THE NUMBER OF CONNECTIONS	GAIN 135
004335	000	400 DO 460 I=1,NMIN2	GAIN 136
004336	000	K=I+1	GAIN 137
004337	000	DO 450 J=K,NFCIL	GAIN 138
004338	000	IF(E2(I).GT.E2(J))GO TO 450	GAIN 139
004339	000	KEEP=E2(I)	GAIN 140
004340	000	LEEP=WORK5(I)	GAIN 141
004341	000	E2(I)=E2(J)	GAIN 142
004342	000	E2(J)=KEEP	GAIN 143
004343	000	WORK5(I)=WORK5(J)	GAIN 144
004344	000	WORK5(J)=LEEP	GAIN 145
004345	000	450 CONTINUE	GAIN 146
004346	000	460 CONTINUE	GAIN 147
004347	000	C	GAIN 148
004348	000	IF(KSNSW(25).EQ.0)GO TO 500	GAIN 149
004349	000	WRITE(6,27)	GAIN 150
004350	000	WRITE(6,18)( E2(I),I=1,NFCIL)	GAIN 151
004351	000	WRITE(6,12)(WORK5(I),I=1,NFCIL)	GAIN 152
004352	000	C FLAG THE ELEMENTS	GAIN 153
004353	000	500 DO 530 I=1,NFCIL	GAIN 154
004354	000	K=WORK4(I)	GAIN 155
004355	000	R1(K)=I	GAIN 156
004356	000	K=WORK5(I)	GAIN 157
004357	000	R2(K)=I	GAIN 158

004358	000	500 CONTINUE	GAIN 159
004359	000	C	GAIN 160
004360	000	IF(KSNSW(26).EQ.0)GO TO 500	GAIN 161
004361	000	WRITE(6,28)	GAIN 162
004362	000	WRITE(6,12)(WORK7(I),I=1,NFCIL)	GAIN 163
004363	000	WRITE(6,29)( R1(I),I=1,NFCIL)	GAIN 164
004364	000	WRITE(6,31)( R2(I),I=1,NFCIL)	GAIN 165
004365	000	C	GAIN 166
004366	000	C	GAIN 167
004367	000	C E)FIND THE COMBINED RANK	GAIN 168
004368	000	500 DO 580 I=1,NFCIL	GAIN 169
004369	000	R(I)=R1(I)+R2(I)	GAIN 170
004370	000	WORK6(I)=I	GAIN 171
004371	000	580 CONTINUE	GAIN 172
004372	000	IF(KSNSW(26).EQ.1)WRITE(6,32)(R(I),I=1,NFCIL)	GAIN 173
004373	000	C	GAIN 174
004374	000	C F)RANK THE ELEMENTS IN AN ASCENDING ORDER OF R	GAIN 175
004375	000	DO 600 I=1,NMIN2	GAIN 176
004376	000	K=I+1	GAIN 177
004377	000	DO 590 J=K,NFCIL	GAIN 178
004378	000	IF(R(I).LT.R(J))GO TO 590	GAIN 179
004379	000	KEEP=R(I)	GAIN 180
004380	000	LEEP=WORK6(I)	GAIN 181
004381	000	R(I)=R(J)	GAIN 182
004382	000	WORK6(I)=WORK6(J)	GAIN 183
004383	000	R(J)=KEEP	GAIN 184
004384	000	WORK6(J)=LEEP	GAIN 185
004385	000	590 CONTINUE	GAIN 186
004386	000	600 CONTINUE	GAIN 187
004387	000	DO 620 I=1,NFCIL	GAIN 188
004388	000	K=WORK6(I)	GAIN 189
004389	000	IFLAG(K)=I	GAIN 190
004390	000	620 CONTINUE	GAIN 191
004391	000	C	GAIN 192
004392	000	IF(KSNSW(27).EQ.0)GO TO 600	GAIN 193
004393	000	WRITE(6,29)( R1(I),I=1,NFCIL)	GAIN 194
004394	000	WRITE(6,31)( R2(I),I=1,NFCIL)	GAIN 195
004395	000	WRITE(6,32)( R(I),I=1,NFCIL)	GAIN 196
004396	000	WRITE(6,07)	GAIN 197
004397	000	WRITE(6,24)(IFLAG(I),I=1,NFCIL)	GAIN 198
004398	000	WRITE(6,36)(WORK6(I),I=1,NFCIL)	GAIN 199
004399	000	C	GAIN 200
004400	000	C	GAIN 201
004401	000	C START ASSIGNING THE ELEMENTS	GAIN 202
004402	000	650 DO 660 I=1,NFCIL	GAIN 203
004403	000	660 FCFLG(I)=.FALSE.	GAIN 204
004404	000	DO 670 J=1,NLOCT	GAIN 205
004405	000	670 FLOGT(J)=.FALSE.	GAIN 206
004406	000	C	GAIN 207
004407	000	C	GAIN 208
004408	000	C PICK THE ELEMENT IN THE BEST RANK , LOCATE IT , THEN PICK THE ONE	GAIN 209
004409	000	C WITH MAXIMUM INTERACTIONS WITH IT , LOCATE IT SO AS TO MINIMIZE THE	GAIN 210
004410	000	C INCREASE IN TOTAL COST . CONTINUE UNTIL ALL ELEMENTS ARE LOCATED	GAIN 211
004411	000	DO 950 I=1,NFCIL	GAIN 212
004412	000	K=WORK6(I)	GAIN 213
004413	000	IF(FCFLG(K))GO TO 950	GAIN 214
004414	000	DO 680 J=1,NLOCT	GAIN 215

004415	000	L=WORK3(J)	GAIN 216
004416	000	IF(FLOGT(L))GO TO 680	GAIN 217
004417	000	LOCAT(K)=L	GAIN 218
004418	000	FCFLG(K)=.TRUE.	GAIN 219
004419	000	FLOGT(L)=.TRUE.	GAIN 220
004420	000	IF(KSN5W(20).EQ.1)WRITE(6,21)K,L	GAIN 221
004421	000	IKHTR=1	
004422	000	IF(I.EQ.1)GO TO 685	
004423	000	KL=K	
004424	000	GO TO 750	GAIN 222
004425	000	680 CONTINUE	GAIN 223
004426	000	WRITE(6,99)K	GAIN 224
004427	000	99 FORMAT(02X,3('*** ELEMENT NO *** '),1 NO PLACE FOR',I6)	GAIN 225
004428	000	GO TO 2000	GAIN 226
004429	000	C PICK THE ELEMENT WITH MAXIMUM INTERACTIONS WITH K	GAIN 227
004430	000	685 KEEP=-10	GAIN 228
004431	000	LEEP=-10	GAIN 229
004432	000	IKHTR=0	
004433	000	DO 690 KK=1,NFCIL	GAIN 230
004434	000	IF(I.EQ.KK)GO TO 690	GAIN 231
004435	000	KL=WORK6(KK)	GAIN 232
004436	000	IF(FCFLG(KL))GO TO 690	GAIN 233
004437	000	ISAV=F(K,KL)	GAIN 234
004438	000	IF(F(KL,K).GT.ISAV)ISAV=F(KL,K)	GAIN 235
004439	000	IF(ISAV.LT.KEEP)GO TO 690	GAIN 236
004440	000	KEEP=ISAV	GAIN 237
004441	000	LEEP=KL	GAIN 238
004442	000	690 CONTINUE	GAIN 239
004443	000	IF(LEEP.LT.0.AND.KSN5W(20).EQ.1)WRITE(6,100)	
004444	000	98 FORMAT(02X,3('*** ELEMENT NO *** '),1 NO MORE ELEMENTS TO CHOSE	GAIN 241
004445	000	*FROM')	GAIN 242
004446	000	IF(LEEP.LT.0)GO TO 700	
004447	000	IF(LEEP.GT.0)GO TO 700	GAIN 243
004448	000	KL=K	GAIN 244
004449	000	GO TO 750	GAIN 245
004450	000	C	GAIN 246
004451	000	C LOCATE THIS ELEMENT SO AS TO RESULT IN MINIMAL INCREASE IN COST	GAIN 247
004452	000	700 IWD=32767	GAIN 248
004453	000	KL=LEEP	GAIN 249
004454	000	DO 710 J=1,NLOCT	GAIN 250
004455	000	L=WORK3(J)	GAIN 251
004456	000	IF(FLOGT(L))GO TO 710	GAIN 252
004457	000	KAM=0	GAIN 253
004458	000	DO 705 II=1,NFCIL	GAIN 254
004459	000	LL=WORK6(II)	GAIN 255
004460	000	IF(LL.EQ.KL)GO TO 705	GAIN 256
004461	000	IF(.NOT.FCFLG(LL))GO TO 705	GAIN 257
004462	000	LAM=LOCAT(LL)	GAIN 258
004463	000	ISAV=F(KL,LL)	GAIN 259
004464	000	IF(F(LL,KL).GT.F(KL,LL))ISAV=F(LL,KL)	GAIN 260
004465	000	KAM=KAM+(ISAV*D(L,LAM))	GAIN 261
004466	000	705 CONTINUE	GAIN 262
004467	000	IF(KAM.GT.IWD)GO TO 710	GAIN 263
004468	000	IWD=KAM	GAIN 264
004469	000	LOCAT(KL)=L	GAIN 265
004470	000	710 CONTINUE	GAIN 266
004471	000	J=LOCAT(KL)	GAIN 267

004472	000	FLOGT(J)=.TRUE.	GAIN 268
004473	000	FCFLG(KL)=.TRUE.	GAIN 269
004474	000	IF(KSNSW(28).EQ.1)WRITE(6,34)KL,J	GAIN 270
004475	000	C	GAIN 271
004476	000	C	GAIN 272
004477	000	C	GAIN 273
004478	000	C	GAIN 274
004479	000	750 KOMP=0	GAIN 275
004480	000	C EVALUATE THIS PARTIAL ASSIGNMENT	GAIN 276
004481	000	DO 770 II=1,NFCIL	GAIN 277
004482	000	LL=WORK6(II)	GAIN 278
004483	000	IF(.NOT.FCFLG(LL))GO TO 770	GAIN 279
004484	000	LAM=LOCAT(LL)	GAIN 280
004485	000	DO 760 KK=1,NFCIL	GAIN 281
004486	000	KM=WORK6(KK)	GAIN 282
004487	000	IF(.NOT.FCFLG(KM))GO TO 760	GAIN 283
004488	000	KAM=LOCAT(KM)	GAIN 284
004489	000	ISAV=F(LL,KM)	GAIN 285
004490	000	IF(ISAV.LT.F(KM,LL))I=AVENUE(LL)	GAIN 286
004491	000	KOMP=KOMP+(ISAV*D(LAM,KAM))	GAIN 287
004492	000	760 CONTINUE	GAIN 288
004493	000	770 CONTINUE	GAIN 289
004494	000	KOMP=0.5*FLOAT(KOMP)	GAIN 290
004495	000	C	GAIN 291
004496	000	IF(KSNSW(29).EQ.1)WRITE(6,46)(LOCAT(II),II=1,NFCIL),KOMP	GAIN 292
004497	000	C	GAIN 293
004498	000	C PERFORM ALL POSSIBLE INSERTIONS AND DELETIONS FOR LAST ELEMENT ONLY	GAIN 294
004499	000	DO 775 II=1,NFCIL	GAIN 295
004500	000	775 RESRV(II)=LOCAT(II)	GAIN 296
004501	000	KOUNT=0	GAIN 297
004502	000	DO 790 II=1,NFCIL	GAIN 298
004503	000	LL=WORK6(II)	GAIN 299
004504	000	IF(LL.EQ.KL)GO TO 790	GAIN 300
004505	000	IF(.NOT.FCFLG(LL))GO TO 790	GAIN 301
004506	000	KOUNT=KOUNT+1	GAIN 302
004507	000	KEY=0	
004508	000	778 KEEJ=LOCAT(KL)	
004509	000	KEEN=LOCAT(LL)	
004510	000	IA=0	
004511	000	DO 785 KR=1,NFCIL	
004512	000	KK=WORK6(KR)	
004513	000	IF(.NOT.FCFLG(KK))GO TO 785	
004514	000	IF(KK.EQ.KL)GO TO 782	
004515	000	IF(F(KK,KL).EQ.0)GO TO 782	
004516	000	LAM=LOCAT(KK)	
004517	000	IA=IA+(F(KK,KL)*D(KEEJ,LAM))	
004518	000	C	
004519	000	782 IF(KK.EQ.LL)GO TO 785	
004520	000	IF(F(KK,LL).EQ.0)GO TO 785	
004521	000	KAM=LOCAT(KK)	
004522	000	IA=IA+(F(KK,LL)*D(KEEN,KAM))	
004523	000	785 CONTINUE	
004524	000	IF(KEY.EQ.1)GO TO 788	
004525	000	IB=IA	
004526	000	KEY=1	
004527	000	KEEP=LOCAT(LL)	GAIN 303
004528	000	LOCAT(LL)=LOCAT(KL)	GAIN 304

004529	000	LOCAT(KL)=KEEP	GAIN 305
004530	000	GO TO 778	
004531	000	788 KEEN= (IA-IB)	
004532	000	KOST(KOUNT)=KOMP+KEEN	
004533	000	DO 786 J=1,NFCIL	GAIN 322
004534	000	786 MAKAN(KOUNT,J)=LOCAT(I)	GAIN 323
004535	000	LEEP=LOCAT(LL)	GAIN 324
004536	000	LOCAT(LL)=KEEP	GAIN 325
004537	000	LOCAT(KL)=LEEP	GAIN 326
004538	000	C	GAIN 327
004539	000	790 CONTINUE	GAIN 328
004540	000	C	GAIN 329
004541	000	C WRITE THE RESULTS OF THE SINGLE INSERTIONS AND DELETIONS TEST IF YOU	GAIN 330
004542	000	C WISK	GAIN 331
004543	000	IF(KSNSW(29),EQ.0)GO TO 800	GAIN 332
004544	000	WRITE(6,37)	GAIN 333
004545	000	DO 810 J=1,KOUNT	GAIN 334
004546	000	810 WRITE(6,38)J,(MAKAN(KOUNT,J)-KOST(J)	GAIN 335
004547	000	38 FORMAT(06X,I3,2X,01,1X,16)	GAIN 336
004548	000	C CHECK FOR IMPROVEMENT	GAIN 337
004549	000	820 LEEP=KOMP	GAIN 338
004550	000	DO 850 J=1,KOUNT	GAIN 339
004551	000	IF(KOST(J).GT.LEEP)GO TO 830	GAIN 340
004552	000	LEEP=KOST(J)	GAIN 341
004553	000	KEEP=J	GAIN 342
004554	000	850 CONTINUE	GAIN 343
004555	000	C	GAIN 344
004556	000	IF(LEEP.EQ.KOMP)GO TO 860	GAIN 345
004557	000	IF(KSNSW(29),EQ.1)WRITE(6,39)LEEP,KOMP	
004558	000	KOMP=LEEP	
004559	000	DO 860 J=1,NFCIL	GAIN 347
004560	000	860 LOCAT(J)=MAKAN(KELP,J)	GAIN 348
004561	000	IF(KSNSW(29),EQ.1)WRITE(6,40)LOCAT(J),J=1,NFCIL	
004562	000	GO TO 900	GAIN 350
004563	000	C	GAIN 351
004564	000	870 IF(KSNSW(29),EQ.1)WRITE(6,41)	
004565	000	DO 880 J=1,NFCIL	GAIN 353
004566	000	880 LOCAT(J)=RESRV(J)	GAIN 354
004567	000	C	GAIN 355
004568	000	C	GAIN 356
004569	000	900 IF(IKHTR.EQ.1)GO TO 920	
004570	000	920 CONTINUE	GAIN 357
004571	000	C	GAIN 358
004572	000	970 WRITE(6,43)KOMP	GAIN 359
004573	000	WRITE(6,44)(LOCAT(I),I=1,NFCIL)	GAIN 360
004574	000	C	GAIN 361
004575	000	C	GAIN 362
004576	000	C	GAIN 363
004577	000	C WRITE THE LOCATIONS OF THE ELEMENTS	GAIN 364
004578	000	WRITE(6,09)	GAIN 365
004579	000	WRITE(6,16)(LOCAT(I),I=1,NFCIL)	GAIN 366
004580	000	C	GAIN 367
004581	000	C	
004582	000	C PERFORM THE TEST OF UNEMPLOYING AND DELETIONS IF DESIRED	
004583	000	IF(IROTA.EQ.0)GO TO 2000	
004584	000	WRITE(6,51)	
004585	000	DO 1090 I=1,NFCIL	



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004586 000 1090 KPLOC(I)=LOCAT(I)
004587 000 KMAX=0
004588 000 ITER=0
004589 000 KADD=0
004590 000 KURCS=KOMP
004591 000 1100 ITER=ITER+1
004592 000 DO 1300 IL=1,NFCIL
004593 000 DO 1105 II=1,NFCIL
004594 000 1105 RESRV(II)=LOCAT(II)
004595 000 ITHSN=0
004596 000 KL=WORK6(IL)
004597 000 KOUNT=0
004598 000 DO 1180 II=1,NFCIL
004599 000 LL=WORK6(II)
004600 000 IF(LL.EQ.KL)GO TO 1140
004601 000 KOUNT=KOUNT+1
004602 000 KEY=0
004603 000 1110 KEEJ=LOCAT(KL)
004604 000 KEEN=LOCAT(LL)
004605 000 IA=0
004606 000 DO 1130 KR=1,NFCIL
004607 000 KK=WORK6(KR)
004608 000 IF(KK.EQ.KL)GO TO 1120
004609 000 IF(F(KK,KL).EQ.0)GO TO 1120
004610 000 LAM=LOCAT(KK)
004611 000 IA=IA+(F(KK,KL)*D(KEEJ,LAM))
004612 000 C
004613 000 C
004614 000 1120 IF(KK.EQ.LL)GO TO 1130
004615 000 IF(F(KK,LL).EQ.0)GO TO 1130
004616 000 KAM=LOCAT(KK)
004617 000 IA=IA+(F(KK,LL)*D(KEEN,KAM))
004618 000 1130 CONTINUE
004619 000 IF(KEY.EQ.1)GO TO 1150
004620 000 IB=IA
004621 000 KEY=1
004622 000 KEEP=LOCAT(LL)
004623 000 LOCAT(LL)=LOCAT(KL)
004624 000 LOCAT(KL)=KEEP
004625 000 GO TO 1110
004626 000 1150 KEEN=(IA-IB)
004627 000 KOST(KOUNT)=KURCS+KEEN
004628 000 DO 1160 J=1,NFCIL
004629 000 1160 MAKAN(KOUNT,J)=LOCAT(J)
004630 000 LEEP=LOCAT(LL)
004631 000 LOCAT(LL)=KEEP
004632 000 LOCAT(KL)=LEEP
004633 000 C
004634 000 1180 CONTINUE
004635 000 C
004636 000 C WRITE THE RESULTS OF THE TEST IF YOU WISH
004637 000 IF(KSNSW(31).EQ.0)GO TO 1220
004638 000 WRITE(6,37)
004639 000 DO 1200 J=1,KOUNT
004640 000 1200 WRITE(6,38)J,(MAKAN(J,KK),KK=1,35),KOST(J)
004641 000 C
004642 000 C CHECK FOR IMPROVEMENTS

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004043 000 1220 LEEP=32767
004044 000 DO 1240 J=1,KOUNT
004045 000 IF(KOST(J).GT.LEEP)GO TO 1230
004046 000 LEEP=KOST(J)
004047 000 KEEPE=J
004048 000 1240 CONTINUE
004049 000 DO 1250 J=1,NFCIL
004050 000 1250 COMPR(IL,J)=MAKAN(KEEP,J)
004051 000 CSTCM(IL)=LEEP
004052 000 C
004053 000 IF(LEEP.GE.KOMP)GO TO 1270
004054 000 IF(KSNSW(32).EQ.1)WRITE(6,97)LEEP,KOMP
004055 000 KOMP=LEEP
004056 000 DO 1260 J=1,NFCIL
004057 000 KPLOC(J)=MAKAN(KOMP,J)
004058 000 1260 LOCAT(J)=MAKAN(KOMP,J)
004059 000 IF(KSNSW(32).EQ.1)WRITE(6,97)LOCAT(J),J=1,NFCIL
004060 000 ITHSN=1
004061 000 KMAX=0
004062 000 GO TO 1500
004063 000 C
004064 000 1270 IF(KSNSW(32).EQ.1)WRITE(6,97)
004065 000 DO 1280 J=1,NFCIL
004066 000 1280 LOCAT(J)=RESRV(J)
004067 000 C
004068 000 1300 CONTINUE
004069 000 C
004070 000 1500 IF(ITHSN.EQ.1.AND.KSNSW(32).EQ.1)WRITE(6,97)
004071 000 KURCS=KOMP
004072 000 IF(ITHSN.EQ.1)GO TO 1100
004073 000 C
004074 000 IF(KSNSW(34).EQ.1)WRITE(6,97)
004075 000 IF(KSNSW(34).EQ.1)WRITE(6,97)LOCAT(I),I=1,NFCIL
004076 000 IF(KSNSW(34).EQ.1)WRITE(6,97)KEEP,KOMP
004077 000 KEEP=-10
004078 000 LEEP=32767
004079 000 DO 1530 I=1,NFCIL
004080 000 IF(CSTCM(I).GE.LEEP)GO TO 1550
004081 000 IF(CSTCM(I).EQ.KOMP)GO TO 1550
004082 000 IF(KADD.EQ.0)GO TO 1550
004083 000 DO 1510 II=1,KADD
004084 000 KEY=0
004085 000 DO 1505 J=1,NFCIL
004086 000 IF(ARRAY(II,J).EQ.COMPR(I,J))KEY=KEY+1
004087 000 1505 CONTINUE
004088 000 IF(KEY.NE.NFCIL)GO TO 1510
004089 000 GO TO 1530
004090 000 1510 CONTINUE
004091 000 1520 LEEP=CSTCM(I)
004092 000 KEEP=I
004093 000 1530 CONTINUE
004094 000 IF(KEEP.GT.0)GO TO 1535
004095 000 WRITE(6,97)
004096 000 97 FORMAT(10X,'HELP ME ,I AM STUCK')
004097 000 GO TO 1800
004098 000 1535 KMAX=KMAX+1
004099 000 IF(KMAX.GT.MAXM)GO TO 1800

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004700	000	KADD=KADD+1	
004701	000	IF(KADD.LE.90)GO TO 1537	
004702	000	WRITE(6,96)	
004703	000	96 FORMAT(10X,'RAN OUT OF ARRAY SPACE')	
004704	000	GO TO 1000	
004705	000	1537 DO 1540 I=1,NFCIL	
004706	000	ARRAY(KADD,I)=COMPR(KEEP,I)	
004707	000	1540 LOCAT(I)=COMPR(KEEP,I)	
004708	000	KURCS=CSTCM(KEEP)	
004709	000	IF(KSNSW(33).EQ.1)WRITE(6,55)LOCAT(I),I=1,NFCIL)	
004710	000	IF(KSNSW(33).EQ.1)WRITE(6,56)KURCS	
004711	000	GO TO 1100	
004712	000	C	
004713	000	C	
004714	000	1800 WRITE(6,9)	
004715	000	WRITE(6,16)(KPLOC(I),I=1,NFCIL)	
004716	000	WRITE(6,53)ITER,KOMP	
004717	000	C	
004718	000	2000 IF(MORE)3000,3000,10	GAIN 368
004719	000	3000 STOP	GAIN 369
004720	000	C	GAIN 401
004721	000	C	GAIN 370
004722	000	1 FORMAT(10X,40I3)	GAIN 371
004723	000	2 FORMAT(10X,3('*** INITIAL'))	GAIN 372
004724	000	4 FORMAT(1H1,' DATA FOR PROBLEM',15/	GAIN 373
004725	000	* 10X,' NUMBER OF ACTIVITIES =',I3/	GAIN 374
004726	000	* 10X,' NUMBER OF LOCATIONS =',I3//	GAIN 375
004727	000	5 FORMAT(10X,' DISTANCE BETWEEN',/)	GAIN 376
004728	000	6 FORMAT(1H1)	GAIN 377
004729	000	7 FORMAT(/)	GAIN 378
004730	000	8 FORMAT(/10X,'INTERACTION MATRIX'/)	GAIN 379
004731	000	9 FORMAT(/10X,'LOCATIONS OF ELEMENTS ARE')	GAIN 380
004732	000	11 FORMAT(' DISTANCES',/)	GAIN 381
004733	000	12 FORMAT(' INDICES',/)	GAIN 382
004734	000	14 FORMAT(/2X,'POSITIONS OF ELEMENTS IN THE RANK',30I3/36X,10I3)	GAIN 383
004735	000	16 FORMAT(' LOCATIONS',/)	GAIN 384
004736	000	17 FORMAT(' NO OF ELEMENTS',/)	GAIN 385
004737	000	18 FORMAT(' NO OF CONNECTIONS',/)	GAIN 386
004738	000	21 FORMAT(1H0,'ARRANGED ELEMENTS AND THEIR INDICES FOLLOW'/)	GAIN 387
004739	000	24 FORMAT(/2X,'POSITIONS OF ELEMENTS IN THE RANK',30I3/36X,10I3)	GAIN 388
004740	000	26 FORMAT(/10X,'ARRAY OF RANKS',11S FOLLOWS'/)	GAIN 389
004741	000	27 FORMAT(/10X,'ARRAY OF RANKS',11S FOLLOWS'/)	GAIN 390
004742	000	28 FORMAT(/10X,'DISPLAY OF RANKS',/)	GAIN 391
004743	000	29 FORMAT(' R1',/)	GAIN 392
004744	000	31 FORMAT(' R2',/)	GAIN 393
004745	000	32 FORMAT(' R',/)	GAIN 394
004746	000	33 FORMAT(10X,'**MAJOR **',/)	GAIN 395
004747	000	34 FORMAT(10X,'**MINOR **',/)	GAIN 396
004748	000	C	GAIN 398
004749	000	36 FORMAT(' CONT OF POSIT',/)	GAIN 397
004750	000	37 FORMAT(15X,'POSSIBLE ASSIGNMENTS')	GAIN 400
004751	000	39 FORMAT(10X,'POSSIBLE IMPROVEMENT', ORIG COST=',I6,' NEW COST=',	GAIN 402
004752	000	*I6,' ASSOCIATED WITH PATERN',I3,' ABOVE')	GAIN 403
004753	000	41 FORMAT(20X,'CORRESPONDING PARTIAL ASSIGNMENT',/20X,35I3)	GAIN 404
004754	000	C	GAIN 399
004755	000	42 FORMAT(10X,'NO IMPROVEMENT POSSIBLE')	GAIN 405
004756	000	43 FORMAT(/10X,'BEST INITIAL SOLUTION S COST=',I6)	GAIN 406

004757	000	44 FORMAT(/15X,'CORRELATION LOCATIONS ARE',5X,'0013)	GAIN 407
004758	000	46 FORMAT(/05X,'ORIGINAL FACILITY ASSIGNMENT',/05X,'0013,04X,'ITS COST=',	GAIN 408
004759	000	* ,I6)	GAIN 409
004760	000	C	
004761	000	51 FORMAT(/10X,'RESULT OF THE OPTIMUM FACILITY ASSIGNMENT')	
004762	000	52 FORMAT(/10X,3(' '),/10X,'THE RESULT IS AN INFEASIBLE SOLUTION,	
004763	000	1WE RESTART AGAIN. IF (NFCIL.EQ.0) GO TO 53	
004764	000	53 FORMAT(/10X,'NUMBER OF FACILITIES =',I2,' COST=',I8)	
004765	000	54 FORMAT(02X,'WE START FROM',5X,'0013)	
004766	000	56 FORMAT(/9X,'WHOSE COST IS',I8)	
004767	000	END	GAIN 411
004768	000	DELTA,SI TUMER-S*SINAH-04JLHNTAID,135627061112	
004769	000	C A SUBROUTINE TO READ THE DATA FOR THE QUADRATIC ASSIGNMENT PROBLEM	GAIN 412
004770	000	C THIS VERSION IS INTENDED FOR THE INGENUITY ET AL DATA	GAIN 413
004771	000	SUBROUTINE DTAIN(JOB,NFCIL,NLOC,NFOT,NFIXD,NRPF,NLOC,ITERM)	GA
004772	000	COMMON IC1,IC2,IC3,IC4,IC5,IC6,IC7,IC8,IC9,IC10,IC11,IC12,IC13,IC14,IC15,IC16,IC17,IC18,IC19,IC20,IC21,IC22,IC23,IC24,IC25,IC26,IC27,IC28,IC29,IC30,IC31,IC32,IC33,IC34,IC35,IC36,IC37,IC38,IC39,IC40,IC41,IC42,IC43,IC44,IC45,IC46,IC47,IC48,IC49,IC50,IC51,IC52,IC53,IC54,IC55,IC56,IC57,IC58,IC59,IC60,IC61,IC62,IC63,IC64,IC65,IC66,IC67,IC68,IC69,IC70,IC71,IC72,IC73,IC74,IC75,IC76,IC77,IC78,IC79,IC80,IC81,IC82,IC83,IC84,IC85,IC86,IC87,IC88,IC89,IC90,IC91,IC92,IC93,IC94,IC95,IC96,IC97,IC98,IC99,IC100,IC101,IC102,IC103,IC104,IC105,IC106,IC107,IC108,IC109,IC110,IC111,IC112,IC113,IC114,IC115,IC116,IC117,IC118,IC119,IC120,IC121,IC122,IC123,IC124,IC125,IC126,IC127,IC128,IC129,IC130,IC131,IC132,IC133,IC134,IC135,IC136,IC137,IC138,IC139,IC140,IC141,IC142,IC143,IC144,IC145,IC146,IC147,IC148,IC149,IC150,IC151,IC152,IC153,IC154,IC155,IC156,IC157,IC158,IC159,IC160,IC161,IC162,IC163,IC164,IC165,IC166,IC167,IC168,IC169,IC170,IC171,IC172,IC173,IC174,IC175,IC176,IC177,IC178,IC179,IC180,IC181,IC182,IC183,IC184,IC185,IC186,IC187,IC188,IC189,IC190,IC191,IC192,IC193,IC194,IC195,IC196,IC197,IC198,IC199,IC200,IC201,IC202,IC203,IC204,IC205,IC206,IC207,IC208,IC209,IC210,IC211,IC212,IC213,IC214,IC215,IC216,IC217,IC218,IC219,IC220,IC221,IC222,IC223,IC224,IC225,IC226,IC227,IC228,IC229,IC230,IC231,IC232,IC233,IC234,IC235,IC236,IC237,IC238,IC239,IC240,IC241,IC242,IC243,IC244,IC245,IC246,IC247,IC248,IC249,IC250,IC251,IC252,IC253,IC254,IC255,IC256,IC257,IC258,IC259,IC260,IC261,IC262,IC263,IC264,IC265,IC266,IC267,IC268,IC269,IC270,IC271,IC272,IC273,IC274,IC275,IC276,IC277,IC278,IC279,IC280,IC281,IC282,IC283,IC284,IC285,IC286,IC287,IC288,IC289,IC290,IC291,IC292,IC293,IC294,IC295,IC296,IC297,IC298,IC299,IC300,IC301,IC302,IC303,IC304,IC305,IC306,IC307,IC308,IC309,IC310,IC311,IC312,IC313,IC314,IC315,IC316,IC317,IC318,IC319,IC320,IC321,IC322,IC323,IC324,IC325,IC326,IC327,IC328,IC329,IC330,IC331,IC332,IC333,IC334,IC335,IC336,IC337,IC338,IC339,IC340,IC341,IC342,IC343,IC344,IC345,IC346,IC347,IC348,IC349,IC350,IC351,IC352,IC353,IC354,IC355,IC356,IC357,IC358,IC359,IC360,IC361,IC362,IC363,IC364,IC365,IC366,IC367,IC368,IC369,IC370,IC371,IC372,IC373,IC374,IC375,IC376,IC377,IC378,IC379,IC380,IC381,IC382,IC383,IC384,IC385,IC386,IC387,IC388,IC389,IC390,IC391,IC392,IC393,IC394,IC395,IC396,IC397,IC398,IC399,IC400,IC401,IC402,IC403,IC404,IC405,IC406,IC407,IC408,IC409,IC410,IC411,IC412,IC413,IC414,IC415,IC416,IC417,IC418,IC419,IC420,IC421,IC422,IC423,IC424,IC425,IC426,IC427,IC428,IC429,IC430,IC431,IC432,IC433,IC434,IC435,IC436,IC437,IC438,IC439,IC440,IC441,IC442,IC443,IC444,IC445,IC446,IC447,IC448,IC449,IC450,IC451,IC452,IC453,IC454,IC455,IC456,IC457,IC458,IC459,IC460,IC461,IC462,IC463,IC464,IC465,IC466,IC467,IC468,IC469,IC470,IC471,IC472,IC473,IC474,IC475,IC476,IC477,IC478,IC479,IC480,IC481,IC482,IC483,IC484,IC485,IC486,IC487,IC488,IC489,IC490,IC491,IC492,IC493,IC494,IC495,IC496,IC497,IC498,IC499,IC500,IC501,IC502,IC503,IC504,IC505,IC506,IC507,IC508,IC509,IC510,IC511,IC512,IC513,IC514,IC515,IC516,IC517,IC518,IC519,IC520,IC521,IC522,IC523,IC524,IC525,IC526,IC527,IC528,IC529,IC530,IC531,IC532,IC533,IC534,IC535,IC536,IC537,IC538,IC539,IC540,IC541,IC542,IC543,IC544,IC545,IC546,IC547,IC548,IC549,IC550,IC551,IC552,IC553,IC554,IC555,IC556,IC557,IC558,IC559,IC560,IC561,IC562,IC563,IC564,IC565,IC566,IC567,IC568,IC569,IC570,IC571,IC572,IC573,IC574,IC575,IC576,IC577,IC578,IC579,IC580,IC581,IC582,IC583,IC584,IC585,IC586,IC587,IC588,IC589,IC590,IC591,IC592,IC593,IC594,IC595,IC596,IC597,IC598,IC599,IC600,IC601,IC602,IC603,IC604,IC605,IC606,IC607,IC608,IC609,IC610,IC611,IC612,IC613,IC614,IC615,IC616,IC617,IC618,IC619,IC620,IC621,IC622,IC623,IC624,IC625,IC626,IC627,IC628,IC629,IC630,IC631,IC632,IC633,IC634,IC635,IC636,IC637,IC638,IC639,IC640,IC641,IC642,IC643,IC644,IC645,IC646,IC647,IC648,IC649,IC650,IC651,IC652,IC653,IC654,IC655,IC656,IC657,IC658,IC659,IC660,IC661,IC662,IC663,IC664,IC665,IC666,IC667,IC668,IC669,IC670,IC671,IC672,IC673,IC674,IC675,IC676,IC677,IC678,IC679,IC680,IC681,IC682,IC683,IC684,IC685,IC686,IC687,IC688,IC689,IC690,IC691,IC692,IC693,IC694,IC695,IC696,IC697,IC698,IC699,IC700,IC701,IC702,IC703,IC704,IC705,IC706,IC707,IC708,IC709,IC710,IC711,IC712,IC713,IC714,IC715,IC716,IC717,IC718,IC719,IC720,IC721,IC722,IC723,IC724,IC725,IC726,IC727,IC728,IC729,IC730,IC731,IC732,IC733,IC734,IC735,IC736,IC737,IC738,IC739,IC740,IC741,IC742,IC743,IC744,IC745,IC746,IC747,IC748,IC749,IC750,IC751,IC752,IC753,IC754,IC755,IC756,IC757,IC758,IC759,IC760,IC761,IC762,IC763,IC764,IC765,IC766,IC767,IC768,IC769,IC770,IC771,IC772,IC773,IC774,IC775,IC776,IC777,IC778,IC779,IC780,IC781,IC782,IC783,IC784,IC785,IC786,IC787,IC788,IC789,IC790,IC791,IC792,IC793,IC794,IC795,IC796,IC797,IC798,IC799,IC800,IC801,IC802,IC803,IC804,IC805,IC806,IC807,IC808,IC809,IC810,IC811,IC812,IC813,IC814,IC815,IC816,IC817,IC818,IC819,IC820,IC821,IC822,IC823,IC824,IC825,IC826,IC827,IC828,IC829,IC830,IC831,IC832,IC833,IC834,IC835,IC836,IC837,IC838,IC839,IC840,IC841,IC842,IC843,IC844,IC845,IC846,IC847,IC848,IC849,IC850,IC851,IC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004814	000	C		GAIN 457
004815	000		DO 60 I=1,NFIXD	GAIN 458
004816	000		60 READ(5,1) FIX(I),WHERE(I)	GAIN 459
004817	000	C		GAIN 460
004818	000		80 DO 999 I=1,60	GAIN 461
004819	000		999 KSNSW(I)=0	GAIN 462
004820	000		DO 998 I=28,40	
004821	000		998 KSNSW(I)=0	
004822	000	C		GAIN 463
004823	000		RETURN	GAIN 464
004824	000	C		GAIN 465
004825	000	C		GAIN 466
004826	000	C		GAIN 467
004827	000		1 FORMAT(20X,12I5)	GAIN 468
004828	000		2 FORMAT(10X,35I2)	GAIN 469
004829	000		END	GAIN 470
004830	000		DELTA,SI TUMER-S*SINAN-QAIN.TIMEOUT,,,152574103711	
004831	000		SUBROUTINE TIMEOUT	
004832	000		K=I*TIME(I,J)	
004833	000		T=I/5000.	
004834	000		WRITE(6,2000)T	
004835	000		2000 FORMAT(' TIME',F10.3,' SEC')	
004836	000		RETURN	
004837	000		END	
004838	000		DELTA,SI TUMER-S*NEWQAP.MAIN,,,152574103711	
004839	000		243	
004840	000	C	A NEW PROGRAM FOR THE GENERAL ASSIGNMENT PROBLEM	
004841	000		COMMON NFCL,NLOC,LEADY,LEADY	
004842	000		COMMON MSTER(05200),TIME(0400),KSNSW(0060)	
004843	000		COMMON NKNAP,KHADY,KHADY,ADRY	
004844	000	C		
004845	000		INTEGER F(20,20),D(20,20),U(20,20),V(20,20),JAS(20,20),ALPHA(20)	
004846	000		LOGICAL Y(20,20),YREST(20,20)	
004847	000		INTEGER WORK1(20,20),WORK2(20,20)	
004848	000		INTEGER UREST(20,20),VREST(20,20)	
004849	000		INTEGER INDEX(0400),INDEX(0400),INDEX(0400)	
004850	000		INTEGER CNVXU(20,20),CNVXU(20,20)	
004851	000	C		
004852	000		EQUIVALENCE( MSTER(0001), F(1,1))	
004853	000		EQUIVALENCE( MSTER(0001), D(1,1))	
004854	000		EQUIVALENCE( MSTER(0001), U(1,1))	
004855	000		EQUIVALENCE( MSTER(1001), V(1,1))	
004856	000		EQUIVALENCE( MSTER(1001), Y(1,1))	
004857	000		EQUIVALENCE( MSTER(2001), F(1,1))	
004858	000		EQUIVALENCE( MSTER(2001), D(1,1))	
004859	000		EQUIVALENCE( MSTER(2001), U(1,1))	
004860	000		EQUIVALENCE( MSTER(3001), V(1,1))	
004861	000		EQUIVALENCE( MSTER(3001), Y(1,1))	
004862	000		EQUIVALENCE( MSTER(4001), F(1,1))	
004863	000		EQUIVALENCE( MSTER(4001), D(1,1))	
004864	000		EQUIVALENCE( MSTER(4001), U(1,1))	
004865	000	C		
004866	000		EQUIVALENCE( IWORK(0001), WORK1(1,1))	
004867	000		EQUIVALENCE( IWORK(0401), WORK2(1,1))	
004868	000		EQUIVALENCE( IWORK(1001), JAS(1,1))	
004869	000		EQUIVALENCE( IWORK(1401), CNVXU(1,1))	
004870	000	C		

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004871 000 C
004872 000 C INITIALIZE VARIOUS COUNTERS AND ARRAYS
004873 000 10 ITER=+1
004874 000 IUBCS=+3276700
004875 000 LBCST=-3276700
004876 000 KOUNT=0
004877 000 C
004878 000 LIMIT=10
004879 000 NKNAP=0
004880 000 KHADY=0
004881 000 KGARY=0
004882 000 KUVW=0
004883 000 C
004884 000 DO 190 I=1,0400
004885 000 190 INDEX(I)=99
004886 000 C
004887 000 C
004888 000 C READ THE KEYS CONTROLLING THE PRINTING IN THE MAIN PROGRAM
004889 000 READ(5,02)(KNSW(I),I=1,100)
004890 000 C READ THE KEYS CONTROLLING THE PRINTING OF U,V,W
004891 000 READ(5,02)(KNSW(I),I=1,100)
004892 000 C
004893 000 C READ THE PROBLEM DATA
004894 000 CALL DTAIN(JOB,LAMDA,ITER,NONE,ITERM,IAVER,ALW)
004895 000 C
004896 000 C PRINT THE DATA IF YOU WISH
004897 000 IF(KNSW(01).EQ.0)GO TO 270
004898 000 WRITE(6,03) JOB,NFCIL,NLOCT
004899 000 WRITE(6,04)LAMDA,ISTART,ITEM
004900 000 270 IF(KNSW(02).EQ.0)GO TO 275
004901 000 WRITE(6,08)
004902 000 DO 275 I=1,NFCIL
004903 000 275 WRITE(6,02)(F(I,J),J=1,NLOCT)
004904 000 WRITE(6,09)
004905 000 DO 280 I=1,NLOCT
004906 000 280 WRITE(6,02)(D(I,J),J=1,NFCIL)
004907 000 WRITE(6,07)
004908 000 C GET THE STARTING VALUES OF U,V,W'S
004909 000 290 CALL UVW(1)
004910 000 DO 293 I=1,NFCIL
004911 000 DO 293 J=1,NLOCT
004912 000 CNVXV(I,J)=0
004913 000 293 CNVXU(I,J)=0
004914 000 AMU=.1*FLOAT(MEW)
004915 000 C
004916 000 C
004917 000 C CALCULATE THE VALUES OF ALPHA
004918 000 DO 298 I=1,NFCIL
004919 000 KEEP=0
004920 000 DO 295 K=1,NFCIL
004921 000 IF(F(I,K).EQ.0)GO TO 295
004922 000 KEEP=KEEP+1
004923 000 295 CONTINUE
004924 000 ALPHA(I)=KEEP
004925 000 298 CONTINUE
004926 000 C
004927 000 C SOLVE THE NFCIL*(NFCIL-1) KNAPSACK PROBLEMS

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004928 000 300 KAM=0
004929 000 KOUNT=KOUNT+1
004930 000 KEEP=0
004931 000 DO 320 I=1,NFCIL
004932 000 DO 310 K=1,NFCIL
004933 000 IF(I.EQ.K)GO TO 310
004934 000 IF(F(I,K).EQ.0)GO TO 310
004935 000 C
004936 000 CALL KNAP(I,K,JEEP,LFEP,KOST)
004937 000 IF(KAM.LT.0400)GO TO 305
004938 000 WRITE(6,98)
004939 000 GO TO 1500
004940 000 305 KAM=KAM+1
004941 000 INDEX(KAM)=I
004942 000 KNDEX(KAM)=K
004943 000 JNDEX(KAM)=JEEP
004944 000 LNDEX(KAM)=LFEP
004945 000 KEEP=KEEP+KOST
004946 000 IF(KSNSW(06).EQ.0)GO TO 310
004947 000 WRITE(6,16)ITER,I,K,I,JEEP,K,LFEP,KOST
004948 000 310 CONTINUE
004949 000 320 CONTINUE
004950 000 IF(KSNSW(06).EQ.1)WRITE(6,17)ITER,KEEP
004951 000 C
004952 000 C SOLVE THE ASSIGNMENT PROBLEM
004953 000 C A)CONSTRUCT THE COST MATRIX
004954 000 IF(KSNSW(07).EQ.1)WRITE(6,18)ITER
004955 000 DO 350 I=1,NFCIL
004956 000 DO 340 J=1,NLOCT
004957 000 340 WORK1(I,J)=-ALPHA(I)*(U(I,J)+V(I,J))
004958 000 IF(KSNSW(07).EQ.1)WRITE(6,19)WORK1(I,J),J=1,NLOCT
004959 000 350 CONTINUE
004960 000 C
004961 000 C B)CALL THE ASSIGNMENT ROUTINE
004962 000 KEY=1
004963 000 K1=30
004964 000 360 CALL HADY(NFCIL,K1,NROW1,NCOL1,ICOST,KEY)
004965 000 GO TO(380,410),KEY
004966 000 380 CALL GARY(NFCIL,K1,NROW1,NCOL1,KEY)
004967 000 GO TO(360,390),KEY
004968 000 390 WRITE(6,99)
004969 000 GO TO 1500
004970 000 C
004971 000 410 THETA=KEEP+ICOST
004972 000 THETA=THETA/2000000.
004973 000 IF(KSNSW(11).EQ.1)WRITE(6,11)ITER,KEEP,ICOST,THETA
004974 000 IF(THETA.LT.LBCST)GO TO 450
004975 000 LBCST=THETA
004976 000 KOUNT=0
004977 000 C
004978 000 C STORE THE BEST VALUES OF THE U'S AND V'S
004979 000 DO 420 I=1,NFCIL
004980 000 DO 420 J=1,NLOCT
004981 000 UBEST(I,J)=U(I,J)
004982 000 420 VBEST(I,J)=V(I,J)
004983 000 C
004984 000 C

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004985 000 C **EVALUATE THE PRIMAL SOLUTION OUT OF THE ASSIGNMENT ALGORITHM
004986 000 C A) INITIALIZE THE Y'S
004987 000 430 DO 435 I=1,NFCIL
004988 000 DO 435 J=1,NLOCT
004989 000 Y(I,J)=.FALSE.
004990 000 435 CONTINUE
004991 000 C
004992 000 C
004993 000 C B) CALCULATE THE QUADRATIC COST
004994 000 KEEP=IUBCS
004995 000 CALL UVW(2)
004996 000 IF(KNSW(12).EQ.1)WRITE(6,12)IF2,IUBCS,KEEP
004997 000 C
004998 000 IF(IUBCS.GE.KEEP)GO TO 480
004999 000 C C) CALCULATE THE BEST SOLUTION VECTOR
005000 000 DO 470 I=1,NFCIL
005001 000 DO 465 J=1,NLOCT
005002 000 YBEST(I,J)=.FALSE.
005003 000 IF(Y(I,J))YBEST(I,J)=.TRUE.
005004 000 465 CONTINUE
005005 000 470 CONTINUE
005006 000 GO TO 490
005007 000 480 IUBCS=KEEP
005008 000 490 IF(LBCST.GE.(IUBCS-151000))GO TO 900
005009 000 C
005010 000 C
005011 000 KEY=0
005012 000 C **UPDATE THE VALUES OF U'S,V'S AND W'S
005013 000 C A) CALCULATE THE U'S
005014 000 IT=0
005015 000 IF(KNSW(16).EQ.1)WRITE(6,20)
005016 000 DO 540 I=1,NFCIL
005017 000 DO 530 J=1,NLOCT
005018 000 KEEP=0
005019 000 DO 515 K=1,NFCIL
005020 000 IF(I.EQ.K)GO TO 515
005021 000 IF(F(I,K).EQ.0)GO TO 515
005022 000 DO 514 L=1,NLOCT
005023 000 IF(J.EQ.L)GO TO 514
005024 000 KEEP=KEEP+X(I,J,K,L)
005025 000 514 CONTINUE
005026 000 515 CONTINUE
005027 000 KAM=0
005028 000 IF(Y(I,J))KAM=ALPHA(I)
005029 000 KEEP=KEEP-KAM
005030 000 IF(KEEP.NE.0)KEY=1
005031 000 IF(KNSW(16).EQ.1)WRITE(6,20)I,J,KEEP
005032 000 KEPA=IABS(KEEP)
005033 000 IT=IT+KEPA
005034 000 WORK1(I,J)=KEEP
005035 000 530 CONTINUE
005036 000 540 CONTINUE
005037 000 C
005038 000 C B) UPDATE THE V'S
005039 000 DO 560 K=1,NFCIL
005040 000 DO 570 L=1,NLOCT
005041 000 KEEP=0

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005042 000 DO 555 I=1,NFCIL
005043 000 IF(K.EQ.1)GO TO 555
005044 000 IF(F(K,I).EQ.0)GO TO 555
005045 000 DO 554 J=1,NLOCT
005046 000 IF(J.EQ.1)GO TO 554
005047 000 KEEP=KEEP+X(I,J,K,L)
005048 000 554 CONTINUE
005049 000 555 CONTINUE
005050 000 KAM=0
005051 000 IF(Y(K,L))KAM=ALPHA(K)
005052 000 KEEP=KEEP-KAM
005053 000 IF(KSNSW(16).EQ.1)WRITE(6,29)K,L,KEEP
005054 000 KEPA=IABS(KEEP)
005055 000 IT=IT+KEPA
005056 000 WORK2(K,L)=KEEP
005057 000 IF(KEEP.NE.0)KEY=1
005058 000 570 CONTINUE
005059 000 580 CONTINUE
005060 000 IF(IT.EQ.0)GO TO 800
005061 000 IT=1*IT
005062 000 DO 600 I=1,NFCIL
005063 000 DO 600 J=1,NLOCT
005064 000 ADD=1000000.*FLOAT(WORK1(I,J))*LAMDA/FLOAT(IT)
005065 000 IF(IAVER.EQ.0)GO TO 595
005066 000 T=AMU*FLOAT(WORK1(I,J))*(1-AMU)*CNVXU(I,J)
005067 000 ADD=1000000.*T*LAMDA/FLOAT(IT)
005068 000 CNVXU(I,J)=T
005069 000 595 KDD=ADD+.5
005070 000 U(I,J)=U(I,J)+KDD
005071 000 600 CONTINUE
005072 000 DO 610 K=1,NFCIL
005073 000 DO 610 L=1,NLOCT
005074 000 ADD=1000000.*FLOAT(WORK2(K,L))*LAMDA/FLOAT(IT)
005075 000 IF(IAVER.EQ.0)GO TO 605
005076 000 T=AMU*FLOAT(WORK2(K,L))*(1-AMU)*CNVXV(K,L)
005077 000 ADD=1000000.*T*LAMDA/FLOAT(IT)
005078 000 CNVXV(K,L)=T
005079 000 605 KDD=ADD+.5
005080 000 V(K,L)=V(K,L)+KDD
005081 000 610 CONTINUE
005082 000 C
005083 000 C )PRINT THE U'S AND V'S IF YOU WISH
005084 000 IF(KSNSW(14).EQ.0)GO TO 620
005085 000 WRITE(6,14)
005086 000 DO 620 I=1,NFCIL
005087 000 620 WRITE(6,01)(U(I,J),J=1,NLOCT)
005088 000 WRITE(6,07)
005089 000 DO 630 I=1,NFCIL
005090 000 630 WRITE(6,01)(V(I,J),J=1,NLOCT)
005091 000 C
005092 000 C **TEST WHETHER PRIMAL FEASIBILITY IS ACHIEVED
005093 000 640 IF(KEY.EQ.0)GO TO 800
005094 000 IF(KOUNT.LT.ITER)GO TO 600
005095 000 WRITE(6,23)ITER,LBCST,IUNCS
005096 000 GO TO 1000
005097 000 C
005098 000 C

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000099      000      600 ITER=ITER+1
000100      000      IF(ITER.LT.LIMIT)GO TO 605
000101      000      LIMIT=LIMIT+35
000102      000      ALAMDE=LAMDA
000103      000      ALAMD=.5*ALAMD
000104      000      LAMDA=ALAMD*.5
000105      000      WRITE(6,31)LAMDA
000106      000      31 FORMAT(10X,'LAMDA=','I5')
000107      000      IF(LAMDA.LT.1)LAMDA=1
000108      000      605 IF(ITER.GT.200)GO TO 800
000109      000      GO TO 300
000110      000      C
000111      000      670 WRITE(6,27)ITER,LBCST,IUBCS
000112      000      GO TO 1000
000113      000      C
000114      000      C
000115      000      800 WRITE(6,21)ITER
000116      000      WRITE(6,22)IUBCS
000117      000      GO TO 1000
000118      000      C
000119      000      900 WRITE(6,24)ITER,LBCST,IUBCS,IUBCS
000120      000      C
000121      000      C PRINT THE VALUES OF THE UNCONSTRAINED V'S
000122      000      1000 WRITE(6,14)
000123      000      DO 1010 I=1,NFCIL
000124      000      1010 WRITE(6,01)(UBEST(I,J),J=1,NFCIL)
000125      000      WRITE(6,07)
000126      000      C
000127      000      DO 1020 I=1,NFCIL
000128      000      1020 WRITE(6,01)(VBEST(I,J),J=1,NFCIL)
000129      000      WRITE(6,07)
000130      000      C
000131      000      C
000132      000      C
000133      000      C EDIT THE BEST SOLUTION TO THE ASSIGNMENT PROBLEM
000134      000      1100 WRITE(6,26)
000135      000      DO 1120 I=1,NFCIL
000136      000      DO 1110 J=1,NLOCY
000137      000      JBAS(I,J)=0
000138      000      IF(VBEST(I,J))JBAS(1,J)=1
000139      000      1110 CONTINUE
000140      000      1120 WRITE(6,02)(JBAS(1,J),J=1,NLOCY)
000141      000      C
000142      000      C
000143      000      1500 IF(MORE)2000,2000,10
000144      000      C
000145      000      2000 STOP
000146      000      C
000147      000      C
000148      000      1 FORMAT(10X,'I5I0')
000149      000      2 FORMAT(10X,'40I3')
000150      000      3 FORMAT(10X,'DATA FOR PROBLEM NO.','I5/
000151      000      *      10X,' NO OF FACT PLANTS ','I5/
000152      000      *      10X,' NO OF LOCATIONS ','I5//')
000153      000      4 FORMAT(10X,' VALUE OF FLOW ','I5/
000154      000      *      10X,' TOL FOR CONVERGENCE ','I5/
000155      000      *      10X,' UB ON ITERATIONS ','I5//')

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005213	000	EQUIVALENCE( MSTER(0001),V(1,1))	
005214	000	EQUIVALENCE( MSTER(0001),V(1,1))	
005215	000	EQUIVALENCE( MSTER(0001),V(1,1))	
005216	000	EQUIVALENCE( MSTER(0001),V(1,1))	
005217	000	C	
005218	000	EQUIVALENCE( MSTER(0001),V(1,1))	
005219	000	EQUIVALENCE( MSTER(0001),V(1,1))	
005220	000	EQUIVALENCE( MSTER(0001),V(1,1))	
005221	000	EQUIVALENCE( MSTER(0001),V(1,1))	
005222	000	C	
005223	000	EQUIVALENCE( MSTER(0001),Y(1,1))	
005224	000	EQUIVALENCE( MSTER(0001),V(1,1))	
005225	000	C	
005226	000	C	
005227	000	C	
005228	000	UPDATE CALLS COUNTERS	CAP7 439
005229	000	KUVW=KUVW+1	CAP7 440
005230	000	K1=30	
005231	000	C	
005232	000	IF(KSNSW(21).EQ.1.AND,KAR.EQ.1)WRITE(6,231)	CAP7 442
005233	000	C	
005234	000	C	
005235	000	GO TO (130,782),KAR	CAP7 446
005236	000	150 NMIN1=NFCIL-1	CAP7 447
005237	000	NMIN2=NLOCT-1	
005238	000	C	
005239	000	C	
005240	000	DO 190 I=1,NFCIL	CAP7 449
005241	000	DO 165 K=1,NFCIL	CAP7 450
005242	000	WORK3(K)=F(I,K)	CAP7 453
005243	000	WORK4(K)=K	CAP7 454
005244	000	165 CONTINUE	CAP7 455
005245	000	DO 180 J=1,NMIN1	CAP7 456
005246	000	K=J+1	CAP7 457
005247	000	DO 170 L=K,NFCIL	CAP7 458
005248	000	IF(WORK3(J).GT.WORK3(L))GO TO 170	CAP7 459
005249	000	KEEP=WORK3(J)	CAP7 460
005250	000	WORK3(J)=WORK3(L)	CAP7 461
005251	000	WORK3(L)=KEEP	CAP7 462
005252	000	KEEP=WORK4(J)	CAP7 463
005253	000	WORK4(J)=WORK4(L)	CAP7 464
005254	000	WORK4(L)=KEEP	CAP7 465
005255	000	170 CONTINUE	CAP7 466
005256	000	160 CONTINUE	CAP7 467
005257	000	DO 185 J=1,NFCIL	CAP7 468
005258	000	185 INTR1(I,J)=WORK4(J)	CAP7 469
005259	000	IF(KSNSW(23).EQ.1)WRITE(6,111I,(WORK3(L),L=1,NFCIL)	CAP7 470
005260	000	190 CONTINUE	CAP7 471
005261	000	C	CAP7 472
005262	000	C	CAP7 473
005263	000	C	CAP7 474
005264	000	DO 230 JJ=1,NLOCT	CAP7 475
005265	000	DO 192 K=1,NLOCT	CAP7 476
005266	000	WORK6(K)=0(JJ,K)	CAP7 477
005267	000	IF(WORK6(K).EQ.0)WORK3(K)=9999	CAP7 478
005268	000	WORK4(K)=K	CAP7 479
005269	000	192 CONTINUE	CAP7 480
		C	CAP7 481
			CAP7 482
			CAP7 483
			CAP7 484
			CAP7 485

005270	000	DO 210 J=1,NMIN2	QAP7 486
005271	000	K=J+1	QAP7 487
005272	000	DO 200 L=K,NLOCT	QAP7 488
005273	000	IF(WORK6(J).LT.WORK6(L))GO TO 200	QAP7 489
005274	000	KEEP=WORK6(J)	QAP7 490
005275	000	WORK6(J)=WORK6(L)	QAP7 491
005276	000	WORK6(L)=KEEP	QAP7 492
005277	000	KEEP=WORK4(J)	QAP7 493
005278	000	WORK4(J)=WORK4(L)	QAP7 494
005279	000	WORK4(L)=KEEP	QAP7 495
005280	000	200 CONTINUE	QAP7 496
005281	000	210 CONTINUE	QAP7 497
005282	000	IF(KSNSW(23).EQ.1)WRITE(6,12)JJ,(WORK6(L),L=1,NLOCT)	QAP7 498
005283	000	DO 220 K=1,NLOCT	QAP7 499
005284	000	WORK2(JJ,K)=WORK6(K)	QAP7 500
005285	000	220 DISTI(JJ,K)=WORK4(K)	QAP7 501
005286	000	230 CONTINUE	QAP7 502
005287	000	C	QAP7 503
005288	000	C	QAP7 504
005289	000	C CONSTRUCT THR WORK1 MATRIX	QAP7 505
005290	000	222 DO 320 I=1,NFCIL	QAP7 506
005291	000	KEEP=0	QAP7 508
005292	000	DO 250 L=1,NFCIL	QAP7 509
005293	000	LL=INTRI(I,L)	QAP7 510
005294	000	KEEP=KEEP+1	QAP7 512
005295	000	WORK3(KEEP)=F(I,LL)	QAP7 513
005296	000	250 CONTINUE	QAP7 514
005297	000	C	QAP7 516
005298	000	DO 290 J=1,NLOCT	QAP7 517
005299	000	YBEST(I,J)=.FALSE.	
005300	000	Y(I,J)=.FALSE.	
005301	000	KEEP=0	QAP7 520
005302	000	DO 260 L=1,NLOCT	QAP7 521
005303	000	LL=DISTI(J,L)	QAP7 522
005304	000	KEEP=KEEP+1	QAP7 524
005305	000	WORK4(KEEP)=D(J,LL)	QAP7 525
005306	000	260 CONTINUE	QAP7 526
005307	000	KEEP=0	QAP7 527
005308	000	C HERE WE CALCULATE WHAT IS EQUIVALENT TO AN OLD 103 ELEMENT	QAP7 529
005309	000	DO 270 L=1,NFCIL	
005310	000	270 KEEP=KEEP+(WORK3(L)*WORK4(L))	QAP7 531
005311	000	C	QAP7 539
005312	000	WORK1(I,J)=KEEP	
005313	000	290 CONTINUE	QAP7 543
005314	000	C	QAP7 545
005315	000	320 CONTINUE	QAP7 550
005316	000	C	QAP7 551
005317	000	C CALCULATE THE REDUCTIONS	QAP7 552
005318	000	KEEP=0	QAP7 553
005319	000	DO 350 I=1,NFCIL	QAP7 554
005320	000	WORK3(I)=32767	QAP7 555
005321	000	DO 330 J=1,NLOCT	QAP7 556
005322	000	IF(WORK1(I,J).GE.WORK3(I))GO TO 330	QAP7 557
005323	000	WORK3(I)=WORK1(I,J)	QAP7 558
005324	000	330 CONTINUE	QAP7 559
005325	000	KEEP=KEEP+WORK3(I)	QAP7 560
005326	000	DO 340 J=1,NLOCT	QAP7 561

005327	000	WORK2(I,J)=WORK1(I,J)	0AP7 562
005328	000	340 WORK1(I,J)=WORK1(I,J)-WORK3(I)	0AP7 563
005329	000	350 CONTINUE	0AP7 564
005330	000	C	0AP7 565
005331	000	DO 400 J=1,NLOCT	0AP7 566
005332	000	WORK6(J)=0.0	0AP7 567
005333	000	DO 375 I=1,NFCIL	0AP7 568
005334	000	IF(WORK1(I,J).GE.WORK6(J).GO TO 375	0AP7 569
005335	000	WORK6(J)=WORK1(I,J)	0AP7 570
005336	000	375 CONTINUE	0AP7 571
005337	000	DO 380 I=1,NFCIL	0AP7 572
005338	000	380 WORK1(I,J)=WORK1(I,J)-WORK6(J)	0AP7 573
005339	000	400 CONTINUE	0AP7 574
005340	000	IF(NFCIL.EQ.NLOCT)GO TO 405	0AP7 575
005341	000	DO 405 I=1,10	0AP7 576
005342	000	405 WRITE(6,99)	0AP7 577
005343	000	99 FORMAT(10X,2(' *** VARIATION *** '),2X,' IN ORDER TO ARRANGE AN	0AP7 578
005344	000	* ARRAY IN LUN02 SO THAT THE SUBROUTINE	0AP7 579
005345	000	STOP	0AP7 580
005346	000	410 DO 415 J=1,NFCIL	0AP7 581
005347	000	415 KEEP=KEEP+WORK6(J)	0AP7 582
005348	000	LBCST=KEEP	0AP7 583
005349	000	C	0AP7 584
005350	000	C PRINT THE MATRICES IF YOU WISH	0AP7 585
005351	000	IF(KSNSW(24).EQ.0)GO TO 420	0AP7 586
005352	000	WORK4(1)=0.0	0AP7 587
005353	000	DO 420 J=2,40	0AP7 588
005354	000	420 WORK4(J)=J-1	0AP7 589
005355	000	WRITE(6,16)	0AP7 590
005356	000	WRITE(6,01)(WORK4(J),J=1,40)	0AP7 591
005357	000	WRITE(6,07)	0AP7 592
005358	000	DO 430 I=1,NFCIL	0AP7 593
005359	000	430 WRITE(6,01)I,(WORK2(I,J),J=1,NLOCT),WORK1(I)	0AP7 594
005360	000	WRITE(6,07)	0AP7 595
005361	000	WRITE(6,01)NFCIL,(WORK6(J),J=1,NLOCT),KEEP	0AP7 596
005362	000	WRITE(6,07)	0AP7 597
005363	000	WRITE(6,18)	0AP7 598
005364	000	WRITE(6,01)(WORK4(J),J=1,40)	0AP7 599
005365	000	WRITE(6,07)	0AP7 600
005366	000	DO 440 I=1,NFCIL	0AP7 601
005367	000	440 WRITE(6,01)I,(WORK1(I,J),J=1,NLOCT),WORK2(I)	0AP7 602
005368	000	WRITE(6,07)	0AP7 603
005369	000	WRITE(6,01)NFCIL,(WORK6(J),J=1,NLOCT),KEEP	0AP7 604
005370	000	WRITE(6,07)	0AP7 605
005371	000	C	0AP7 606
005372	000	450 IF(KSNSW(25).EQ.1)WRITE(6,19)KEEP	0AP7 607
005373	000	C	
005374	000	ITOT=0	0AP7 613
005375	000	C	
005376	000	C ASSIGN THE INITIAL VALUES OF THE U'S AND V'S	
005377	000	K=0	
005378	000	DO 480 I=1,NFCIL	
005379	000	K=0	
005380	000	DO 480 J=1,NLOCT	
005381	000	K=K+1	
005382	000	U(I,J)=-0*1000000	
005383	000	V(I,J)=-0*1000000	

005384	000	UBEST(I,J)=U(I,J)	
005385	000	VBEST(I,J)=V(I,J)	
005386	000	480 CONTINUE	
005387	000	C	QAP7 621
005388	000	C INITIALIZE HADY	QAP7 622
005389	000	750 N1=NFCIL	QAP7 623
005390	000	IF(NFCIL.EQ.NLOCT)GO TO 760	QAP7 624
005391	000	C ADD ADUMMY FACILITY	QAP7 625
005392	000	N1=NFCIL+1	QAP7 626
005393	000	DO 755 J=1,NLOCT	QAP7 627
005394	000	755 WORK1(N1,J)=0	QAP7 628
005395	000	IORIG(N1)=NLOCT-NFCIL	QAP7 629
005396	000	760 DO 765 I=1,NFCIL	QAP7 630
005397	000	765 IORIG(I)=1	QAP7 631
005398	000	DO 770 J=1,NLOCT	QAP7 632
005399	000	770 IDEST(J)=1	QAP7 633
005400	000	KEY=2	QAP7 634
005401	000	773 CALL HADY(N1,K1,NLOCT,HCPL1,TTOT,KEY)	QAP7 635
005402	000	GO TO(775,781),KEY	
005403	000	775 CALL GARY(N1,K1,HCOWF,HCPL1,KEY)	QAP7 637
005404	000	GO TO(773,780),KEY	QAP7 638
005405	000	780 WRITE(6,99)	QAP7 639
005406	000	STOP	
005407	000	99 FORMAT(02X,3(' ***** ERROR IN GARY ***'))	QAP7 641
005408	000	781 MBCST=ITOT+LBCST	
005409	000	IF(KSNSW(28).EQ.1)WRITE(6,23)ITOT	QAP7 644
005410	000	C	QAP7 645
005411	000	IC2=0.5*FLOAT(MBCST)	QAP7 646
005412	000	KEY=1	QAP7 647
005413	000	C EVALUATE THE SOLUTION	QAP7 649
005414	000	782 DO 784 K=1,NLOCT	
005415	000	784 WORK8(K)=0	QAP7 651
005416	000	LTOT=0	
005417	000	DO 790 I=1,NFCIL	QAP7 652
005418	000	DO 785 K=1,NLOCT	QAP7 653
005419	000	IF(WORK8(K).EQ.1)GO TO 785	QAP7 654
005420	000	IF(UBAS(I,K).LE.K1)GO TO 785	QAP7 655
005421	000	WORK4(I)=K	QAP7 656
005422	000	Y(I,K)=.TRUE.	
005423	000	IF(KAM.EQ.1)YBEST(I,K)=.TRUE.	QAP7 657
005424	000	WORK8(K)=1	QAP7 658
005425	000	GO TO 790	QAP7 659
005426	000	785 CONTINUE	QAP7 660
005427	000	790 CONTINUE	QAP7 661
005428	000	C	
005429	000	IF(KSNSW(30).EQ.0)GO TO 798	
005430	000	DO 796 I=1,NFCIL	
005431	000	DO 795 J=1,NLOCT	
005432	000	WORK2(I,J)=0	
005433	000	IF(Y(I,J))WORK2(I,J)=1	
005434	000	795 CONTINUE	
005435	000	WRITE(6,01)(WORK2(I,J),J=1,NLOCT)	
005436	000	796 CONTINUE	
005437	000	798 DO 810 I=1,NFCIL	QAP7 662
005438	000	DO 800 J=1,NFCIL	QAP7 663
005439	000	LOCI=WORK4(I)	QAP7 664
005440	000	LOCJ=WORK4(J)	QAP7 665

[illegible]

005496	000	C		QAP71126
005499	000		DO 30 I=1,NFCIL	QAP71127
005500	000		DO 25 J=1,NFCIL	QAP71128
005501	000		IF(I.EQ.J)GO TO 25	QAP71129
005502	000		IF(I.GT.J)F(I,J)=WORK(I,J)	QAP71130
005503	000		IF(I.LT.J)O(I,J)=WORK(I,J)	QAP71131
005504	000	25	CONTINUE	QAP71132
005505	000	30	CONTINUE	QAP71133
005506	000	C		QAP71134
005507	000		DO 40 I=1,NFCIL	QAP71135
005508	000		DO 40 J=1,NFCIL	QAP71136
005509	000		F(I,J)=F(J,I)	QAP71137
005510	000		D(J,I)=D(I,J)	QAP71138
005511	000	40	CONTINUE	QAP71139
005512	000		DO 50 I=1,30	
005513	000	50	KNSW(I)=1	
005514	000	C		QAP71140
005515	000	C		QAP71141
005516	000		KNSW(1)=1	QAP71142
005517	000		KNSW(4)=1	QAP71143
005518	000		KNSW(6)=0	QAP71144
005519	000		KNSW(07)=0	QAP71145
005520	000		KNSW(16)=0	
005521	000		KNSW(14)=0	
005522	000		KNSW(08)=1	QAP71146
005523	000		KNSW(30)=0	
005524	000		KNSW(25)=0	
005525	000		KNSW(02)=0	
005526	000		KNSW(23)=0	
005527	000		KNSW(24)=0	
005528	000	C	READ THE INDICES OF THE FACILITIES FIXED APRIORI	QAP71147
005529	000		DO 70 I=1,NFIXD	QAP71148
005530	000	70	READ(5,1) FIX(I),WHERE(I)	QAP71149
005531	000	C		QAP71150
005532	000		NFIXD=1	QAP71151
005533	000		RETURN	QAP71152
005534	000	C		QAP71153
005535	000	C		QAP71154
005536	000	C		QAP71155
005537	000		1 FORMAT(20X,12I5)	QAP71156
005538	000		2 FORMAT(10X,35I2)	QAP71157
005539	000		END	
005540	000		DELTA,SI TUMER-S*NEWQAP.KNAP,,,13773080112	
005541	000	C	A SUBROUTINE TO SOLVE THE KNAPSACK PROBLEM	
005542	000		SUBROUTINE KNAP(I,K,J,NFCIL,KAM)	
005543	000		COMMON NFCIL,NLOC,LLOC,LUNCS	
005544	000		COMMON MSTER(05200),IWORK(0000),KNSW(0060)	
005545	000		COMMON NKNAP,KHADY,KCARY,NECTP	
005546	000	C		
005547	000	C		
005548	000		INTEGER U(20,20),V(20,20)	
005549	000		INTEGER F(20,20),D(20,20)	
005550	000		EQUIVALENCE ( MSTER(0001), F(1,1))	
005551	000		EQUIVALENCE ( MSTER(0401), D(1,1))	
005552	000	C		
005553	000		EQUIVALENCE (MSTER(0001), U(1,1))	
005554	000		EQUIVALENCE (MSTER(1201), V(1,1))	



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005555 000 C
005556 000 C UPDATE THE COUNTER BY 1
005557 000 NKNAP=NKNAP+1
005558 000 KAM=327670000
005559 000 DO 130 J=1,NLOCT
005560 000 DO 120 L=1,NLOCT
005561 000 IF(J.EQ.L)GO TO 120
005562 000 KOMP=(1000000*(F(I,K)+D(I,J)+U(I,J)+V(K,L))
005563 000 IF(KSN5W(06).EQ.1.AND.2*(L-2)*WRITE(6,11))I,J,K,L,KOMP
005564 000 11 FORMAT(30X,'C(',415,'75411)
005565 000 IF(KOMP.GT,KAM)GO TO 120
005566 000 KAM=KOMP
005567 000 JEEP=J
005568 000 LEEP=L
005569 000 C
005570 000 120 CONTINUE
005571 000 130 CONTINUE
005572 000 C
005573 000 RETURN
005574 000 C
005575 000 END
005576 000 QFLT,SI TUMER=5*NEWQAP.HQY,1-10057612
005577 000 C *****CAP7 718
005578 000 C A SUBROUTINE TO FIND AN OPTIMAL SOLUTION OF A TRANSPORTATION PROBLEM CAP7 719
005579 000 C USING THE METHOD OF MODIFIED GRADIENTS CAP7 720
005580 000 C APPLICABLE TO DEGENERATE PROBLEMS BUT NOT TO UNDETERMINED ONES CAP7 721
005581 000 C ALL X'S HAVE TO BE MODIFIED BY SUBTRACTION CAP7 722
005582 000 C <<+++++*****CAP7 723
005583 000 C CAP7 724
005584 000 SUBROUTINE HADY(NH010,75411,NCOL1,ITOT,KEY) CAP7 725
005585 000 C -----CAP7 726
005586 000 COMMON NCOL,NDEST,IND,ICOST CAP7 727
005587 000 COMMON MSTER(05200),IWORK(1000),KSN5W(06) CAP7 728
005588 000 COMMON NKNAP,KHADY,REAL,NDECTP CAP7 729
005589 000 C CAP7 730
005590 000 DIMENSION ICOST(20,20),IOWR(20),IDEST(20),QJAS(20,20)
005591 000 DIMENSION IU(20),IV(20),IU1(20),IV1(20)
005592 000 C CAP7 733
005593 000 C CAP7 734
005594 000 EQUIVALENCE( IWORK(001),ICOST(1,1)) CAP7 735
005595 000 EQUIVALENCE( IWORK(002),ICOST(001)) CAP7 736
005596 000 EQUIVALENCE( IWORK(003),IDEST(001)) CAP7 737
005597 000 EQUIVALENCE( IWORK(004), IU(001)) CAP7 738
005598 000 EQUIVALENCE( IWORK(005), IV(001)) CAP7 739
005599 000 EQUIVALENCE( IWORK(006), IU1(001)) CAP7 740
005600 000 EQUIVALENCE( IWORK(007), IV1(001)) CAP7 741
005601 000 EQUIVALENCE( IWORK(100), QJAS(1,1)) CAP7 742
005602 000 C <<+++++*****CAP7 743
005603 000 C UPDATE CALLS COUNTER CAP7 744
005604 000 KHADY=KHADY+1 CAP7 745
005605 000 C CAP7 746
005606 000 GO TO(320,5),KEY CAP7 747
005607 000 C -----CAP7 748
005608 000 C INITIALIZE CAP7 749
005609 000 5 K1=30 CAP7 750
005610 000 K3=32767000
005611 000 NDST1=NDEST CAP7 751

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005612	000	NORG1=NORIG	QAP7 752
005613	000	NCOL=1	QAP7 753
005614	000	NROW=1	QAP7 754
005615	000	ISAV=K3	QAP7 755
005616	000	DO 130 I=1,NORIG	QAP7 756
005617	000	DO 130 J=1,NDEST	QAP7 757
005618	000	130 JBAS(I,J)=0	QAP7 758
005619	000	DO 140 J=1,NDEST	QAP7 759
005620	000	140 IV(J)=1	QAP7 760
005621	000	C FROM HERE UNTIL JUST BEFORE 300 THE VARIABLES ARE ALLOCATED	QAP7 761
005622	000	150 IF(ICOST(NROW,NCOL)-ISAV)150,160,160	QAP7 762
005623	000	160 NCOL=NCOL+1	QAP7 763
005624	000	IF(NCOL-NDEST)150,150,200	QAP7 764
005625	000	180 IF(IV(NCOL)-1)160,190,160	QAP7 765
005626	000	190 ISAV=ICOST(NROW,NCOL)	QAP7 766
005627	000	NCOL1=NCOL	QAP7 767
005628	000	GO TO 160	QAP7 768
005629	000	C	QAP7 769
005630	000	200 IF(IORIG(NROW)-IDEST(NCOL1))210,230,310	QAP7 770
005631	000	210 ISET=IORIG(NROW)+K1	QAP7 771
005632	000	C	QAP7 772
005633	000	JBAS(NROW,NCOL1)=ISET	QAP7 773
005634	000	C	QAP7 774
005635	000	IDEST(NCOL1)=IDEST(NCOL1)-IORIG(NROW)	QAP7 775
005636	000	GO TO 290	QAP7 776
005637	000	C	QAP7 777
005638	000	230 ISET=IORIG(NROW)+K1	QAP7 778
005639	000	C	QAP7 779
005640	000	JBAS(NROW,NCOL1)=ISET	QAP7 780
005641	000	C	QAP7 781
005642	000	IV(NCOL1)=0	QAP7 782
005643	000	C MAKE NECESSARY ARRANGEMENTS FOR DEGENERACY	QAP7 783
005644	000	NCOL=1	QAP7 784
005645	000	ISAV=K3	QAP7 785
005646	000	IF(NROW-NORIG)240,320,320	QAP7 786
005647	000	240 IF(ICOST(NROW,NCOL)-ISAV)250,270,270	QAP7 787
005648	000	250 IF(IV(NCOL)-1)270,260,270	QAP7 788
005649	000	260 ISAV=ICOST(NROW,NCOL)	QAP7 789
005650	000	NCOL1=NCOL	QAP7 790
005651	000	270 NCOL=NCOL+1	QAP7 791
005652	000	IF(NCOL-NDEST)240,240,260	QAP7 792
005653	000	C	QAP7 793
005654	000	280 JBAS(NROW,NCOL1)=K1	QAP7 794
005655	000	C	QAP7 795
005656	000	290 NROW=NROW+1	QAP7 796
005657	000	IF(NROW-NORIG)300,300,320	QAP7 797
005658	000	300 NCOL=1	QAP7 798
005659	000	ISAV=K3	QAP7 799
005660	000	GO TO 150	QAP7 800
005661	000	C	QAP7 801
005662	000	310 ISET=IDEST(NCOL1)+K1	QAP7 802
005663	000	C	QAP7 803
005664	000	JBAS(NROW,NCOL1)=ISET	QAP7 804
005665	000	C	QAP7 805
005666	000	IORIG(NROW)=IORIG(NROW)-IDEST(NCOL1)	QAP7 806
005667	000	IV(NCOL1)=0	QAP7 807
005668	000	GO TO 300	QAP7 808

005669	000	C	<<+++++GAP7 809
005670	000	C	START CALCULATING THE DUAL VARIABLES GAP7 810
005671	000	C	-----GAP7 811
005672	000	C	A) INITIALIZE AS FOLLOWS GAP7 812
005673	000	C	* SET ALL DUAL VARIABLES TO A IV EQUAL TO ZERO GAP7 813
005674	000	C	* FLAG ALL DUAL VARIABLES, THEIR IV1 ARE NOT EQUAL TO ZERO GAP7 814
005675	000		320 DO 330 I=1,NORIG GAP7 815
005676	000		IU(I)=0 GAP7 816
005677	000		330 IU1(I)=0 GAP7 817
005678	000		DO 340 J=1,NDEST GAP7 818
005679	000		IV(J)=0 GAP7 819
005680	000		340 IV1(J)=0 GAP7 820
005681	000	C	UNFLAG THE FIRST ROW GAP7 821
005682	000		IU1(1)=1 GAP7 822
005683	000		NROW=1 GAP7 823
005684	000		NCOL=1 GAP7 824
005685	000	C	B) FROM HERE UNTIL JUST BEFORE 540 THE REMAINING DUAL VARIABLES ARE GAP7 825
005686	000	C	DETERMINED BY REFLECTING GAP7 826
005687	000		360 IF (JBAS(NROW,NCOL)-R1) 360,360,370 GAP7 827
005688	000		370 IV(NCOL)=ICOST(NROW,NCOL)-IV(NROW) GAP7 828
005689	000	C	* UNFLAG COLUMN NCOL GAP7 829
005690	000		IV1(NCOL)=1 GAP7 830
005691	000		380 NCOL=NCOL+1 GAP7 831
005692	000		IF (NCOL-NDEST) 360,360,390 GAP7 832
005693	000		400 NROW=NROW+1 GAP7 833
005694	000		IF (NROW-NORIG) 420,420,430 GAP7 834
005695	000		420 IF (IU1(NROW)-1) 430,430,440 GAP7 835
005696	000		430 NCOL=1 GAP7 836
005697	000		440 IF (JBAS(NROW,NCOL)-R1) 440,470,470 GAP7 837
005698	000		450 NCOL=NCOL+1 GAP7 838
005699	000		IF (NCOL-NDEST) 440,450,460 GAP7 839
005700	000		470 IF (IV1(NCOL)) 480,480,490 GAP7 840
005701	000		480 IU(NROW)=ICOST(NROW,NCOL)-IV(NCOL) GAP7 841
005702	000	C	* UNFLAG ROW NROW GAP7 842
005703	000		IU1(NROW)=1 GAP7 843
005704	000		NCOL=1 GAP7 844
005705	000		GO TO 360 GAP7 845
005706	000	C	* CHECK WHETHER ANY ROWS ARE STILL FLAGGED, IF SO RETURN TO 430, GAP7 846
005707	000	C	OTHERWISE GO TO 540 GAP7 847
005708	000		500 NROW=1 GAP7 848
005709	000		510 IF (IU1(NROW)) 520,430,520 GAP7 849
005710	000		520 NROW=NROW+1 GAP7 850
005711	000		IF (NROW-NORIG) 510,510,530 GAP7 851
005712	000	C	<<+++++GAP7 852
005713	000	C	CHECK OPTIMALITY OF THE CURRENT SOLUTION GAP7 853
005714	000	C	-----GAP7 854
005715	000		540 ISAV=0 GAP7 855
005716	000		NROW=1 GAP7 856
005717	000		550 NCOL=1 GAP7 857
005718	000		560 IF (JBAS(NROW,NCOL)-R1) 560,570,570 GAP7 858
005719	000		570 NCOL=NCOL+1 GAP7 859
005720	000		IF (NCOL-NDEST) 560,560,580 GAP7 860
005721	000		590 NROW=NROW+1 GAP7 861
005722	000		IF (NROW-NORIG) 550,550,600 GAP7 862
005723	000		610 IF (ISAV) 700,800,700 GAP7 863
005724	000		620 IS1=IU(NROW)+IV(NCOL)-ICOST(NROW,NCOL) GAP7 864
005725	000		IF (IS1-ISAV) 570,570,640 GAP7 865

005720	000	640 ISAV=IS1	GAP7 866
005727	000	C	GAP7 867
005728	000	NROW1=NROW	GAP7 868
005729	000	NCOL1=NCOL	GAP7 869
005730	000	GO TO 570	GAP7 870
005731	000	C <<+++++*****	GAP7 871
005732	000	C CURRENT SOLUTION IS NON-OPTIMAL	GAP7 872
005733	000	C -----	GAP7 873
005734	000	700 KEY=1	GAP7 874
005735	000	RETURN	GAP7 875
005736	000	C <<+++++*****	GAP7 876
005737	000	C CURRENT SOLUTION IS OPTIMAL	GAP7 877
005738	000	C -----	GAP7 878
005739	000	800 KEY=2	GAP7 879
005740	000	C CALCULATE ACTUAL VALUES OF BASIC VARIABLES	GAP7 880
005741	000	ITOT=0	GAP7 881
005742	000	DO 850 I=1,NORIG	GAP7 882
005743	000	DO 830 J=1,NDEST	GAP7 883
005744	000	IF(JBAS(I,J))830,830,830	GAP7 884
005745	000	820 JBASE =JBAS(I,J)-R1	GAP7 885
005746	000	ITOT=ITOT+(JBASE *ICOL(I,J))	GAP7 886
005747	000	830 CONTINUE	GAP7 887
005748	000	850 CONTINUE	GAP7 888
005749	000	IF(KSNSW(30).EQ.0)GO TO 900	GAP7 889
005750	000	WRITE(6,11)	GAP7 890
005751	000	DO 870 I=1,NORIG	GAP7 891
005752	000	870 WRITE(6,01)I,(JBAS(I,J),J=1,NDEST)	GAP7 892
005753	000	C	GAP7 893
005754	000	900 RETURN	GAP7 894
005755	000	C	GAP7 895
005756	000	1 FORMAT(10X,I3,2X,2013)	GAP7 896
005757	000	11 FORMAT(10X,'OPTIMAL SOLUTION OF THE LINEAR ASSIGNMENT PROBLEM 1/1)	GAP7 897
005758	000	C	GAP7 898
005759	000	C	GAP7 899
005760	000	C <<+++++*****	GAP7 900
005761	000	END	GAP7 901
005762	000	0FLT,SI TUMER-S*NEWGAP,GARY,..10000057612	
005763	000	C	GAP7 902
005764	000	C	GAP7 903
005765	000	C	GAP7 904
005766	000	C *****	GAP7 905
005767	000	C A SUBROUTINE TO IDENTIFY ITS LOOP AND UPDATE ITS ENTRIES	GAP7 906
005768	000	C <<+++++*****	GAP7 907
005769	000	SUBROUTINE GARY(NORIG,K1,NDEST,NCOL1,KEY)	GAP7 908
005770	000	C -----	GAP7 909
005771	000	COMMON NRCIL,NDEST,LRCS,ITOT	GAP7 910
005772	000	COMMON MSTER(05200),IWORK(1000),KSNSW(60)	GAP7 911
005773	000	COMMON NKNAP,KHADY,KGARY,NKCTP	GAP7 912
005774	000	C	GAP7 913
005775	000	DIMENSION JBAS(20,20)	
005776	000	DIMENSION INET(080),INET1(20),INET2(20)	
005777	000	C	GAP7 916
005778	000	EQUIVALENCE( IWORK(0001),INET1(001))	GAP7 917
005779	000	EQUIVALENCE( IWORK(0001),INET2(001))	GAP7 918
005780	000	EQUIVALENCE( IWORK(0001),INET(001))	GAP7 919
005781	000	EQUIVALENCE( IWORK(1001),JBAS(1,1))	GAP7 920
005782	000	C	GAP7 921

005783	000	C		GAP7 922
005784	000	C	<<+++++	GAP7 923
005785	000	C		GAP7 924
005786	000	C		GAP7 925
005787	000	C	UPDATE CALLS COUNTER	GAP7 926
005788	000		KGARY=KGARY+1	GAP7 927
005789	000	C		GAP7 928
005790	000	C	A) INITIALZE	GAP7 929
005791	000	C	* SET K2 .GE. (10*K1)	GAP7 930
005792	000		K2=300000	GAP7 931
005793	000		K=NORIG+NDEST	GAP7 932
005794	000		K=2+K	GAP7 933
005795	000		DO 20 I=1,K	GAP7 934
005796	000		20 INET(I)=0	GAP7 935
005797	000	C	** THUS THE SIZE OF ARRAY INET IS 2*(NORIG-NDEST+1)**	GAP7 936
005798	000		DO 40 I=1,NORIG	GAP7 937
005799	000		40 INET(I)=0	GAP7 938
005800	000		DO 60 I=1,NDEST	GAP7 939
005801	000		60 INET2(I)=0	GAP7 940
005802	000	C	<<+++++	GAP7 941
005803	000	C	B) START SEARCHING FOR THE BASIS LOOP	GAP7 942
005804	000	C	-----	GAP7 943
005805	000	C	** FROM HERE UNTIL LINE 900 GAP 943-948 ARE	GAP7 944
005806	000	C	1) RECOGNIZING THE BASIS OF LOOP AND DETERMINING ITS ROW AND	GAP7 945
005807	000	C	COLUMN NUMBER AND THE ADJACENT LOCATION OF ARRAY INET	GAP7 946
005808	000	C	2) DETECT ANY BASIS ELEMENTS	GAP7 947
005809	000	C	-----	GAP7 948
005810	000		I=1	GAP7 949
005811	000		INET(I)=NROW1	GAP7 950
005812	000		I1=I+1	GAP7 951
005813	000		INET(I1)=NCOL1	GAP7 952
005814	000		NROW=NROW1	GAP7 953
005815	000		NCOL=1	GAP7 954
005816	000		I=I+2	GAP7 955
005817	000		100 IF (JBAS(NROW,NCOL)-K1)100,100,120	GAP7 956
005818	000		120 IF (NCOL-NCOL1)140,100,100	GAP7 957
005819	000		140 INET(I)=NROW	GAP7 958
005820	000		I1=I+1	GAP7 959
005821	000		INET(I1)=NCOL	GAP7 960
005822	000		I=I+2	GAP7 961
005823	000		GO TO 220	GAP7 962
005824	000	C		GAP7 963
005825	000		100 NCOL=NCOL+1	GAP7 964
005826	000		IF (NCOL-NDEST)100,100,200	GAP7 965
005827	000		200 I=1	GAP7 966
005828	000	C	*) NO BASIS ELEMENT IN THE ROW OF THE ENTERING , ERROR TYPE 1	GAP7 967
005829	000		WRITE(6,1)I,NROW1,NCOL1	GAP7 968
005830	000		KEY=2	GAP7 969
005831	000	C		GAP7 970
005832	000		RETURN	GAP7 971
005833	000	C	<<+++++	GAP7 972
005834	000		220 INET2(NCOL)=1	GAP7 973
005835	000		NROW=1	GAP7 974
005836	000		240 IF (JBAS(NROW,NCOL)-K1)240,300,300	GAP7 975
005837	000		200 NROW=NROW+1	GAP7 976
005838	000		IF (NROW-NORIG)240,240,300	GAP7 977
005839	000		300 I=I-2	GAP7 978

005840	000	IF(I)360,360,340	QAP7 979
005841	000	340 NROW=INET(I)	QAP7 980
005842	000	I1=I+1	QAP7 981
005843	000	NCOL=INET(I1)	QAP7 982
005844	000	INET2(NCOL)=0	QAP7 983
005845	000	C	QAP7 984
005846	000	GO TO 520	QAP7 985
005847	000	C *) THERE IS NO BASIS LOOP	QAP7 986
005848	000	360 I=2	QAP7 987
005849	000	WRITE(6,2)I,NROW1,NCOL1	QAP7 988
005850	000	KEY=2	QAP7 990
005851	000	C	QAP7 991
005852	000	RETURN	QAP7 992
005853	000	C <<+++++>>	QAP7 993
005854	000	380 I2=I-2	QAP7 994
005855	000	IF(NROW-INET(I2))400,360,400	QAP7 995
005856	000	400 IF(INET1(NROW))300,420,300	QAP7 996
005857	000	420 INET(I)=NROW	QAP7 997
005858	000	I1=I+1	QAP7 998
005859	000	INET(I1)=NCOL	QAP7 999
005860	000	I=I+2	QAP71000
005861	000	IF(NROW-NROW1)480,460,480	QAP71001
005862	000	480 I=I-2	QAP71002
005863	000	GO TO 300	QAP71003
005864	000	480 INET1(NROW)=1	QAP71004
005865	000	NCOL=1	QAP71005
005866	000	500 IF(JBAS(NROW,NCOL)-K1)520,500,560	QAP71006
005867	000	520 NCOL=NCOL+1	QAP71007
005868	000	IF(NCOL-NREST)500,500,640	QAP71008
005869	000	560 I1=I-1	QAP71009
005870	000	IF(NCOL-INET(I1))580,560,640	QAP71010
005871	000	580 IF(INET2(NCOL))640,600,640	QAP71011
005872	000	600 INET(I)=NROW	QAP71012
005873	000	I1=I+1	QAP71013
005874	000	INET(I1)=NCOL	QAP71014
005875	000	I=I+2	QAP71015
005876	000	IF(NCOL-NCOL1)220,700,220	QAP71016
005877	000	640 I=I-2	QAP71017
005878	000	IF(I)360,360,680	QAP71018
005879	000	680 NROW=INET(I)	QAP71019
005880	000	I1=I+1	QAP71020
005881	000	NCOL=INET(I1)	QAP71021
005882	000	INET1(NROW)=0	QAP71022
005883	000	GO TO 260	QAP71023
005884	000	C	QAP71024
005885	000	700 I=3	QAP71025
005886	000	C *) CALCULATE THE MINIMUM ENTRY IN BASIS LOOP	QAP71026
005887	000	ISAV=K2	QAP71027
005888	000	720 NROW=INET(I)	QAP71028
005889	000	I1=I+1	QAP71029
005890	000	NCOL=INET(I1)	QAP71030
005891	000	IF(JBAS(NROW,NCOL)-ISAV)760,780,780	QAP71031
005892	000	760 ISAV=JBAS(NROW,NCOL)	QAP71032
005893	000	NROW2=NROW	QAP71033
005894	000	NCOL2=NCOL	QAP71034
005895	000	780 IF(NCOL-NCOL1)800,860,800	QAP71035
005896	000	800 I=I+4	QAP71036

005897	000	IF(I-K)720,720,800	GAP71037
005898	000	C *) THE BASIS LOOP HAS MORE ELEMENTS THAN THERE ARE IN THE BASIS .	GAP71038
005899	000	C ERROR TYPE 3	GAP71039
005900	000	840 I=3	GAP71040
005901	000	C	GAP71041
005902	000	WRITE(6,3)I,NROW1,NROW2	GAP71042
005903	000	KEY=2	GAP71043
005904	000	C	GAP71044
005905	000	RETURN	GAP71045
005906	000	C <<+++++>>	GAP71046
005907	000	840 IF(ISAV-K2)900,880,800	GAP71047
005908	000	C *) THERE IS NO ELEMENT IN THE BASIS LOOP JLI. KP: ERROR TYPE 4	GAP71048
005909	000	860 I=4	GAP71049
005910	000	C	GAP71050
005911	000	WRITE(6,4)I,NROW1,NROW2	GAP71051
005912	000	KEY=2	GAP71052
005913	000	C	GAP71053
005914	000	RETURN	GAP71054
005915	000	C <<+++++>>	GAP71055
005916	000	C C) NOW THE LOCATIONS OF ALL ELEMENTS IN THE BASIS ARE SEQUENTIALLY	GAP71056
005917	000	C ARRANGED (PAIRWISE) IN A SORTED ORDER AND EACH CONTAINS THE MINIMUM	GAP71057
005918	000	C ENTRY IN THE LOOP. THE BASIS IS UPDATED FOR EACH SOLUTION BY	GAP71058
005919	000	C SUCCESSIVE ADDITIONS AND SUBTRACTIONS FROM THE ELEMENTS OF THE	GAP71059
005920	000	C BASIS LOOP	GAP71060
005921	000	C -----	GAP71061
005922	000	900 J=-1	GAP71062
005923	000	NROW=INET(1)	GAP71063
005924	000	NCOL=INET(2)	GAP71064
005925	000	JBAS(NROW,NCOL)=ISAV	GAP71065
005926	000	ISAV=ISAV-K1	GAP71066
005927	000	C	GAP71067
005928	000	I=3	GAP71068
005929	000	920 NROW=INET(I)	GAP71069
005930	000	I1=I+1	GAP71070
005931	000	NCOL=INET(I1)	GAP71071
005932	000	IF(NROW-NROW2)1000,1010,1020	GAP71072
005933	000	980 IF(NCOL-NCOL2)1000,1010,1020	GAP71073
005934	000	980 JBAS(NROW,NCOL)=0	GAP71074
005935	000	GO TO 1020	GAP71075
005936	000	C	GAP71076
005937	000	1000 ISET=JBAS(NROW,NCOL)	GAP71077
005938	000	JBAS(NROW,NCOL)=ISET	GAP71078
005939	000	C	GAP71079
005940	000	1020 J=-J	GAP71080
005941	000	I=I+2	GAP71081
005942	000	IF(NCOL-NCOL1)920,1010,1020	GAP71082
005943	000	C <<+++++>>	GAP71083
005944	000	C NOW THAT THE NEW BASIS ELEMENT HAS BEEN OBTAINED, SET KEY = 1 TO	GAP71084
005945	000	C ENABLE HADY TO CHECK FOR CIRCULARITY	GAP71085
005946	000	C -----	GAP71086
005947	000	1080 KEY=1	GAP71087
005948	000	RETURN	GAP71088
005949	000	C	GAP71089
005950	000	1 FORMAT(10X,'ERROR TYPE 1/15X,'NO BASIS ELEMENT IN ROW WHICH CONTAINS	GAP71090
005951	000	*INS ENTERING ELEMENT IN AN ITERATION')	GAP71091
005952	000	2 FORMAT(10X,'ERROR TYPE 2/15X,'THERE IS NO BASIS LOOP',3I3)	GAP71092
005953	000	3 FORMAT(10X,'ERROR TYPE 3/15X,'THE BASIS LOOP HAS MORE ELEMENTS THAN	GAP71093

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005954 000 *AN THERE ARE IN THE BASIS (1313) QAP71094
005955 000 4 FORMAT(10X,'ERROR TYPE 4/ 1 X, 'THERE IS NO ELEMENT IN THE BASIS LOAP71095
005956 000 *OOP LESS THSN K2',313) QAP71096
005957 000 C QAP71097
005958 000 C <<+++++ QAP71098
005959 000 END QAP71099
005960 000 DELT,SI TUMER-S*NEWQAP.X,.,130110061312
005961 000 C A FUNCTION TO CALCULATE THE VALUE OF X
005962 000 FUNCTION X(I,J,K,L)
005963 000 COMMON NFOIL,NLOC1,LOC2,LOC3
005964 000 COMMON MSTER(05200),ILOC(0,000),KSNSW(0060)
005965 000 COMMON NKNAP,KHADY,KCARY,LOC4
005966 000 C
005967 000 INTEGER INDEX(0400),KINDEX(0400),JINDEX(0400),LINDEX(0400)
005968 000 C
005969 000 C
005970 000 EQUIVALENCE (MSTER(0001),INDEX(001))
005971 000 EQUIVALENCE (MSTER(0002),INDEX(001))
005972 000 EQUIVALENCE (MSTER(0003),INDEX(001))
005973 000 EQUIVALENCE (MSTER(0004),INDEX(001))
005974 000 C
005975 000 C
005976 000 X=0
005977 000 KAM=0
005978 000 20 KAM=KAM+1
005979 000 IF (INDEX(KAM).EQ.99)RETURN
005980 000 IF (I.EQ.INDEX(KAM).AND.J.EQ.JINDEX(KAM))GO TO 50
005981 000 GO TO 20
005982 000 50 IF (K.EQ.KINDEX(KAM).AND.L.EQ.LINDEX(KAM))GO TO 90
005983 000 GO TO 20
005984 000 C
005985 000 90 X=1
005986 000 C
005987 000 RETURN
005988 000 C
005989 000 END
005990 000 DELT,SI TUMER-S*QAP-DATA.PROBLEM1,,,130110061312
005991 000 PROBLEM 4001 5 5 70 01 -2 05 9999
005992 000 1 1 0001010203
005993 000 1 2 0500020102
005994 000 1 3 0203000102
005995 000 1 4 0400000001
005996 000 1 5 0102000500
005997 000 1 1
005998 000 DELT,SI TUMER-S*QAP-DATA.PROBLEM2,,,130262061312
005999 000 PROBLEM 4002 6 6 70 01 -2 8 9999 6
006000 000 2 1 000102010203
006001 000 2 2 050001020102
006002 000 2 3 020300030201
006003 000 2 4 040000000102
006004 000 2 5 010200050001
006005 000 2 6 000200021000
006006 000 5 2
006007 000 DELT,SI TUMER-S*QAP-DATA.PROBLEM3,,,120420061312
006008 000 PROBLEM 4003 7 7 70 2 -2 10 9999
006009 000 3 1 00010203020304
006010 000 3 2 05000102010203

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[illegible]



[illegible]

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006182 000 NUM=(NPTS*NPTS/2)-NPTS+2
006183 000 CALL RANDU(X,NUM)
006184 000 IJ=1
006185 000 DO 10 I=1,N
006186 000 COST(I,I)=1.0E38
006187 000 L=I+1
006188 000 DO 10 J=L,NPTS
006189 000 IJ=IJ+1
006190 000 IYFL=A+(B-A)*X(IJ)
006191 000 COST(I,J)=IYFL
006192 000 10 COST(J,I)=IYFL
006193 000 COST(NPTS,NPTS)=1.0E38
006194 000 GO TO 30
006195 000 20 CALL IO(IFIN,2,NPTS,COST)
006196 000 30 IF(PRINTM) CALL SHOW(NPTS,COST)
006197 000 CALL TIMEOUT
006198 000 RETURN
006199 000 END
006200 000 DELT,SI TUMER-S*FATHOM.10,2,13,1,0,0,012
006201 000 SUBROUTINE IO(IFILE,IOPTR,A)
006202 000 C* IO READS OR WRITES A UNIT IN A FILE
006203 000 C* IOPT=1 WRITE
006204 000 C* IOPT=2 READ
006205 000 PARAMETER MAXPTS=500
006206 000 DIMENSION A(MAXPTS,MAXPTS)
006207 000 LOGICAL BAD
006208 000 BAD=.FALSE.
006209 000 5 CALL NTRAN(IFILE,10,22)
006210 000 DO 10 I=1,N
006211 000 CALL NTRAN(IFILE,IOPTR,A(I,I),ISTAT,22)
006212 000 IF(ISTAT.LE.0)GO TO 700
006213 000 10 CONTINUE
006214 000 RETURN
006215 000 700 IF(BAD)GO TO 800
006216 000 BAD=.TRUE.
006217 000 GO TO 5
006218 000 800 WRITE(6,2000)IFILE
006219 000 2000 FORMAT(' FILE ERROR ON UNIT',I3)
006220 000 STOP
006221 000 END
006222 000 DELT,SI TUMER-S*FATHOM.MAIN,1,1,0,0,073212
006223 000 C* THIS PROGRAM SOLVES THE TRAVELING SALESMAN PROBLEM
006224 000 C* BY FATHOMING
006225 000 INCLUDE BLANK
006226 000 DIMENSION PISAVE(20),IPOINT(20)
006227 000 DIMENSION IPOINT(MAXPTS)
006228 000 DIMENSION ACOST(MAXPTS),ICOST(MAXPTS),ITEMP(MAXPTS)
006229 000 DIMENSION ASAVE(MAXPTS),ISAVE(MAXPTS),XSAVE(MAXPTS,MAXPTS)
006230 000 LOGICAL EXSAVE,PRINTL,PRINTM,PRINTV,ALP
006231 000 C* READ DATA
006232 000 READ(5,1000)NPTS,ISTART,CLAVE,EPS,MAXIT,NPI,IIN,L,M,K,MATGEN
006233 000 1000 FORMAT()
006234 000 IF(NPI.EQ.0)GO TO 2
006235 000 READ(5,1000) (IFOUT(I),I=1,NPI)
006236 000 READ(5,1000) (PISAVE(I),I=1,NPI)
006237 000 C* NPTS=NUMBER OF POINTS
006238 000 C* ISTART=STARTING POINT (SEARCH IF ISTART=0)

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006239 000 C*      CSAVE=INITIAL COST MATRIX
006240 000 C*      EPS=STOPPING CRITERIA AND DIFFERENCE BETWEEN BOUNDS
006241 000 C*      MAXIT=MAXIMUM NUMBER OF ITERATIONS
006242 000 C*      NPI=NUMBER OF POINTS
006243 000 C*      IFIN=INPUT FILE FOR COST MATRIX
006244 000 C*      IFOUT=OUTPUT FILE FOR COST MATRIX
006245 000 C*      PISAVE=VALUES FOR ADDITIONAL COST MATRIX
006246 000 C*      PRINTL=PRINT OPTION FOR LIST AT EACH STEP
006247 000 C*      PRINTM=PRINT OPTION FOR COST MATRIX
006248 000 C*      PRINTV=PRINT OPTION FOR VOLL METHOD
006249 000 C*      MATGEN=COST MATRIX IS GENERATED IN RANDOM MANNER(=1)
006250 000 2 PRINTL=(L.GT.0)
006251 000 PRINTM=(M.GT.0)
006252 000 PRINTV=(K.GT.0)
006253 000 IF(MATGEN.EQ.1) GO TO 3
006254 000 CALL INPUT(NPTS,IFIN,CS,IFOUT,PRINTM)
006255 000 GO TO 4
006256 000 3 CALL INRAND(NPTS,IFIN,CS,PRINTM)
006257 000 C*      FIND STARTING POINT AND START COST MATRIX
006258 000 4 CALL START(ISTART,IFIN,PISAVE,IFOUT,PRINTM,PRINTV)
006259 000 C*      READ COST FACTOR FOR FINDING
006260 000 IF(FOUND) GO TO 10
006261 000 READ(5,1000,END=9000)CPCT
006262 000 CFACT=CPCT/100.0
006263 000 IF(CPCT.LE.0.0)CMAX=1.0
006264 000 IF(CPCT.GT.0.0)CMAX=1.0/CFACT
006265 000 CMIN=-1.0E30
006266 000 FOUND=.FALSE.
006267 000 C*      PLACE STARTING POINT ON LIST
006268 000 5 NPTSIN=1
006269 000 IPOINT(1)=ISTART
006270 000 LAST=ISTART
006271 000 C*      FIND MINIMUM CONNECTION
006272 000 J1=MINPT(ISTART,1)
006273 000 J2=MINPT2(ISTART,J1)
006274 000 C*      FIND MINIMUM TREE
006275 000 TCOST(1)=TREE(1,1)END
006276 000 C*      COMPUTE COST
006277 000 C=TCOST(1)+COST(J1)+COST(J2,ISTART)
006278 000 IF(CMIN.LT.C)CMIN=C
006279 000 IF(PRINTL)CALL OUTPUT(C,CMIN,IPOINT)
006280 000 C*      ADD NEW POINT
006281 000 10 NPTSIN=NPTSIN+1
006282 000 IF(NPTSIN.EQ.NPTS)GO TO 20
006283 000 DO 20 I=1,NPTS
006284 000 20 EX(NPTSIN,I)=.FALSE.
006285 000 IF(NPTSIN.EQ.2)J2=J1
006286 000 ACOST(NPTSIN)=ACOST(NPTSIN-1)+COST(LAST,J2)
006287 000 LAST=J2
006288 000 IPOINT(NPTSIN)=LAST
006289 000 IN(LAST)=.TRUE.
006290 000 C*      FIND MINIMUM CONNECTIONS
006291 000 IF(LAST.NE.J1)GO TO 30
006292 000 25 J1=MINPT(ISTART,1)
006293 000 IF(J1.EQ.0)GO TO 40
006294 000 IF(NPTSIN.NE.1)GO TO 30
006295 000 J2=MINPT2(ISTART,J1)

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006296 000 GO TO 35
006297 000 30 J2=MINPT(LAST,NPTSIN)
006298 000 35 IF(J2.EQ.0)GO TO 40
006299 000 C* FIND MINIMUM TREE
006300 000 TCOST(NPTSIN)=TREE(NPTSIN,ITMP)
006301 000 C* COMPUTE COST
006302 000 C=ACOST(NPTSIN)+TCOST(NPTSIN)+COST(J1,ISTART)+COST(LAST,J2)
006303 000 IF(PRINTL)CALL OUTPUT(NPTSIN,C,IPPOINT)
006304 000 IF(C.LE.CMAX)GO TO 10
006305 000 C* REMOVE POINT FROM LIST
006306 000 40 NPTSIN=NPTSIN-1
006307 000 IF(NPTSIN.EQ.0)GO TO 80
006308 000 EX(NPTSIN,LAST)=.TRUE.
006309 000 IN(LAST)=.FALSE.
006310 000 LAST=IPPOINT(NPTSIN)
006311 000 GO TO 25
006312 000 C* COMPUTE COST OF COMPLETE LOOP
006313 000 50 C=ACOST(NPTS-1)+COST(J1,CMIN)+COST(LAST,J2)
006314 000 LAST=J2
006315 000 IPPOINT(NPTS)=LAST
006316 000 IF(CMAX.LT.C)GO TO 40
006317 000 WRITE(6,2000)
006318 000 2000 FORMAT(//,' NEW SOLUTION')
006319 000 CALL OUTPUT(NPTS,C,IPPOINT)
006320 000 CALL TIMEOUT
006321 000 IF(CSAVE.LE.C.AND.FOUND)GO TO 40
006322 000 CSAVE=C
006323 000 IF(CPCT.GT.0.0)GO TO 52
006324 000 C* ADJUST UPPER BOUND
006325 000 CMAX=(C+CMIN)/2.0
006326 000 IF((CMAX-CMIN).LE.EPS)GO TO 90
006327 000 GO TO 55
006328 000 52 CMAX=CFAC*CSAVE
006329 000 55 FOUND=.TRUE.
006330 000 DO 60 I=1,NPTS
006331 000 ISAVE(I)=IPPOINT(I)
006332 000 ASAVE(I)=ACOST(I)
006333 000 TSAVE(I)=TCOST(I)
006334 000 DO 60 J=1,NPTS
006335 000 60 EXSAVE(I,J)=EX(I,J)
006336 000 IF(CMAX.LT.CMIN) GO TO 80
006337 000 GO TO 40
006338 000 C* ADJUST LOWER BOUND
006339 000 80 IF(CPCT.GT.0.0)GO TO 100
006340 000 CMIN=CMAX
006341 000 CALL TIMEOUT
006342 000 CMAX=(CMIN+CSAVE)/2.0
006343 000 IF((CMAX-CMIN).LE.EPS)GO TO 90
006344 000 GO TO 105
006345 000 90 GO TO 101
006346 000 C* PRINT FINAL SOLUTION
006347 000 100 CONTINUE
006348 000 WRITE(6,2100)CPCT
006349 000 2100 FORMAT(//,1X,F5.1,' % SOLUTION')
006350 000 CALL TIMEOUT
006351 000 IF(.NOT.ALK) GO TO 101
006352 000 FOUND=.FALSE.

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006350 000 ALR=.FALSE.
006354 000 CSAVE=1.0F30
006355 000 CPCT=0
006356 000 C* READ COST FACTOR FOR FATHOMING.
006357 000 101 READ(5,1000,END=9000)
006358 000 IF(CPCT1.LE.CPCT) GO TO 101
006359 000 CPCT=CPCT1
006360 000 102 CFACT=CPCT/100.0
006361 000 CMAX=CFACT*CSAVE
006362 000 IF(CMAX.LT.CMIN) GO TO 80
006363 000 105 IF(.NOT.FOUND)GO TO 120
006364 000 DO 110 I=1,NPTS
006365 000 IN(I)=.TRUE.
006366 000 IPOINT(I)=ISAVE(I)
006367 000 ACOST(I)=ASAVE(I)
006368 000 TCOST(I)=TSAVE(I)
006369 000 DO 110 J=1,NPTS
006370 000 110 EX(I,J)=FXSAVE(I,J)
006371 000 LAST=IPOINT(NPTS)
006372 000 NPTSIN=NPTS
006373 000 GO TO 40
006374 000 120 CONTINUE
006375 000 DO 130 I=1,NPTS
006376 000 DO 130 J=1,NPTS
006377 000 130 EX(I,J)=.FALSE.
006378 000 GO TO 5
006379 000 800 WRITE(6,3000)
006380 000 3000 FORMAT(//,' SOLVED ON APPROX')
006381 000 CPCT=100.0
006382 000 CMIN=CMAX
006383 000 FOUND=.FALSE.
006384 000 GO TO 102
006385 000 900 CALL TIMOUT
006386 000 STOP
006387 000 END
006388 000 DELT,SI TUMER=S*FATHOM.MINPT,.,.,13700113111
006389 000 FUNCTION MINPT(I,N)
006390 000 C* MINPT FINDS POINT WITH MINIMUM COST FOR CONNECTION TO
006391 000 C* GIVEN POINT
006392 000 INCLUDE BLANK
006393 000 M=0
006394 000 C=1.0E38
006395 000 DO 10 J=1,NPTS
006396 000 IF(IN(J).OR.EX(N,J))GO TO 10
006397 000 IF(C.LE.COST(I,J))GO TO 10
006398 000 C=COST(I,J)
006399 000 M=J
006400 000 10 CONTINUE
006401 000 MINPT=M
006402 000 RETURN
006403 000 END
006404 000 DELT,SI TUMER=S*FATHOM.MINPT2,.,.,134003113111
006405 000 FUNCTION MINPT2(I,N)
006406 000 C* MINPT2 FINDS POINT WITH SECOND LOWEST COST FOR CONNECTION TO
006407 000 C* GIVEN POINT
006408 000 C* I=GIVEN POINT
006409 000 C* N=POINT WITH LOWEST COST

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006410 000      INCLUDE BLANK
006411 000      M=0
006412 000      C=1.0E38
006413 000      DO 10 J=1,NPTS
006414 000      IF(J.EQ.I.OR.J.EQ.N)GO TO 10
006415 000      IF(EX(1,J))GO TO 10
006416 000      IF(C.LE.COST(I,J))GO TO 10
006417 000      C=COST(I,J)
006418 000      M=J
006419 000      10 CONTINUE
006420 000      MINPT2=M
006421 000      RETURN
006422 000      END
006423 000      DELT,SI TUMER-S*FATHOM.ONLY/VIEW,,,200311121411
006424 000      SUBROUTINE ORDER(ISTART)
006425 000      C*      DUMMY SUBROUTINE
006426 000      RETURN
006427 000      END
006428 000      DELT,SI TUMER-S*FATHOM.ORDER/OLD,,,176251121411
006429 000      SUBROUTINE ORDER(ISTART)
006430 000      C*      ORDER ORDERS LIST OF COMBINATIONS BY INCREASING COST
006431 000      C*      COMBINATIONS CONTAINING STARTING POINT ARE OMITTED
006432 000      INCLUDE BLANK
006433 000      N=NPTS-1
006434 000      NLIST=(N*N-N)/2
006435 000      IF(ISTART.EQ.0)NLIST=(N*N-1)/2
006436 000      DO 30 K=1,NLIST
006437 000      C=1.0E38
006438 000      DO 20 I=1,N
006439 000      IF(I.EQ.ISTART)GO TO 20
006440 000      L=I+1
006441 000      DO 10 J=L,NPTS
006442 000      IF(J.EQ.ISTART.OR.EX(I,J))GO TO 10
006443 000      IF(C.LE.COST(I,J))GO TO 10
006444 000      C=COST(I,J)
006445 000      LIST(1,K)=I
006446 000      LIST(2,K)=J
006447 000      10 CONTINUE
006448 000      20 CONTINUE
006449 000      I=LIST(1,K)
006450 000      J=LIST(2,K)
006451 000      30 EX(I,J)=.TRUE.
006452 000      DO 40 I=1,NPTS
006453 000      DO 40 J=1,NPTS
006454 000      40 EX(I,J)=.FALSE.
006455 000      RETURN
006456 000      END
006457 000      DELT,SI TUMER-S*FATHOM.OUTPUT/BATCH,,,172454113311
006458 000      SUBROUTINE OUTPUT(N,C,IPOINT)
006459 000      C*      OUTPUT PRINTS COST AND LIST OF POINTS
006460 000      DIMENSION IPOINT(1)
006461 000      M=MIN(N,25)
006462 000      WRITE(6,2000)C,(IPOINT(I),I=1,M)
006463 000      IF(N.LE.25)GO TO 900
006464 000      DO 10 L=26,N,25
006465 000      M=MIN(N,L+24)
006466 000      10 WRITE(6,2100) (IPOINT(I),I=L,M)

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006467 000 2000 FORMAT(' COST',F7.0,2X,'END',25I4)
006468 000 2100 FORMAT(22X,25I4)
006469 000 900 RETURN
006470 000 END
006471 000 DELT,SI TUMER-S*FATHOM,START,17261411001
006472 000 SUBROUTINE OUTPUT,PRINTM,PRINTV
006473 000 C* OUTPUT PRINTS COST AND LIST OF POINTS
006474 000 DIMENSION IPOINT(1)
006475 000 M=MIN(N,12)
006476 000 WRITE(6,2000)C,(IPOINT(I),I=1,M)
006477 000 IF(M.LE.12)GO TO 900
006478 000 DO 10 L=13,N,12
006479 000 M=MIN(N,L+11)
006480 000 10 WRITE(6,2100) (IPOINT(I),I=1,M)
006481 000 2000 FORMAT(' COST',F7.0,2X,'END',12I4)
006482 000 2100 FORMAT(22X,12I4)
006483 000 900 RETURN
006484 000 END
006485 000 DELT,SI TUMER-S*FATHOM,START,134704001012
006486 000 SUBROUTINE SHOW(COST,MAXPTS)
006487 000 C* SHOW PRINTS COST MATRIX
006488 000 PARAMETER MAXPTS=80
006489 000 DIMENSION COST(MAXPTS,MAXPTS)
006490 000 WRITE(6,2000)
006491 000 2000 FORMAT(' COST MATRIX')
006492 000 DO 20 K=1,NPTS,16
006493 000 L=MIN(NPTS,K+15)
006494 000 WRITE(6,2100) (J,J=K,L)
006495 000 DO 20 I=1,NPTS
006496 000 20 WRITE(6,2200)I,(COST(I,J),J=K,L)
006497 000 2100 FORMAT(4X,16I7)
006498 000 2200 FORMAT(1X,13,16F7.0)
006499 000 RETURN
006500 000 END
006501 000 DELT,SI TUMER-S*FATHOM,START,135036051012
006502 000 SUBROUTINE SHOW(COST,MAXPTS)
006503 000 C* SHOW PRINTS COST MATRIX
006504 000 PARAMETER MAXPTS=80
006505 000 DIMENSION COST(MAXPTS,MAXPTS)
006506 000 WRITE(6,2000)
006507 000 2000 FORMAT(' COST MATRIX')
006508 000 DO 20 K=1,NPTS,9
006509 000 L=MIN(NPTS,K+8)
006510 000 WRITE(6,2100) (J,J=K,L)
006511 000 DO 20 I=1,NPTS
006512 000 20 WRITE(6,2200)I,(COST(I,J),J=K,L)
006513 000 2100 FORMAT(4X,9I7)
006514 000 2200 FORMAT(1X,13,9F7.0)
006515 000 RETURN
006516 000 END
006517 000 DELT,SI TUMER-S*FATHOM,START,135023070312
006518 000 SUBROUTINE START(ISTART,EXIT,NPT,PISAVE,IFOUT,PRINTM,PRINTV)
006519 000 C* START FINDS THE STARTING POINT AND ADJUSTS THE COST MATRIX
006520 000 INCLUDE BLANK
006521 000 DIMENSION PISAVE(1),IFOUT(1)
006522 000 DIMENSION IPI(MAXPTS),IPEP(MAXPTS),CTEMP(MAXPTS,MAXPTS)
006523 000 LOGICAL PRINTM,PRINTV

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006524 000 C* ORDER CONNECTIONS BY INCREASING COST
006525 000 CALL ORDER(0)
006526 000 C* FIND STARTING POINT WITH MAXIMUM LOWER BOUND OF COST
006527 000 CMAX=-1.0E38
006528 000 JSTART=ISTART
006529 000 DO 20 I=1,NPTS
006530 000 IF(JSTART.NE.0.AND.JSTART.NE.I)GO TO 20
006531 000 IN(I)=.TRUE.
006532 000 T=TREE(1,ITEMP)
006533 000 K1=MINPT(1,1)
006534 000 K2=MINPT2(1,K1)
006535 000 C=T+COST(K1,1)+COST(1,K2)
006536 000 IF(C.LE.CMAX)GO TO 20
006537 000 ISTART=I
006538 000 CMAX=C
006539 000 J1=K1
006540 000 J2=K2
006541 000 DO 10 J=1,NPTS
006542 000 10 IPI(J)=ITEMP(J)
006543 000 20 IN(I)=.FALSE.
006544 000 WRITE(6,2000) CMAX,ISTART
006545 000 2000 FORMAT(' COST',F10.0,' STARTING POINT',13)
006546 000 IN(ISTART)=.TRUE.
006547 000 ISAVE(1)=ISTART
006548 000 C* USE VOGEL METHOD TO FIND SOLUTION
006549 000 IF(CSAVE.LE.0.0)CALL VOGEL(NPTS,COST,PRINTV,CSAVE,ISAVE)
006550 000 IF(NPI.LE.0)GO TO 900
006551 000 ICMIN=NPTS**2
006552 000 CALL COPY(NPTS,COST,CTEMP)
006553 000 CALL TIMEOUT
006554 000 C* ITERATE ON COST MATRIX TO FIND MAXIMUM LOWER BOUND
006555 000 JPI=1
006556 000 PI=PISAVE(1)
006557 000 WRITE(6,2100)PI
006558 000 2100 FORMAT(' PI=',F10.4)
006559 000 N=NPTS-1
006560 000 IT=0
006561 000 C* ADJUST COST
006562 000 25 IPI(ISTART)=0
006563 000 IPI(J1)=IPI(J1)+1
006564 000 IPI(J2)=IPI(J2)+1
006565 000 ICOST=ABS(IPI(NPTS))
006566 000 ICSQ=ICOST**2
006567 000 DO 30 J=1,N
006568 000 ICOST=ICOST+ABS(IPI(J))
006569 000 ICSQ=ICSQ+IPI(J)**2
006570 000 L=J+1
006571 000 DO 30 K=L,NPTS
006572 000 COST(J,K)=COST(J,K)+PI+(IPI(J)+IPI(K))
006573 000 30 COST(K,J)=COST(J,K)
006574 000 IF(ICOST.EQ.0)GO TO 800
006575 000 IF(PRINTM)CALL SHOW(NPTS,COST)
006576 000 C* REORDER CONNECTIONS
006577 000 CALL ORDER(ISTART)
006578 000 C* COMPUTE MINIMUM COST
006579 000 T=TREE(1,IPI)
006580 000 J1=MINPT(ISTART,1)

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006081      000      J2=MINPT2(ISTART,J1)
006082      000      C=I+COST(J1,ISTART)+COST(ISTART,J2)
006083      000      IF(C.LE.CMAX)GO TO 80
006084      000      C*      NEW MAXIMUM FOUND
006085      000      IT=0
006086      000      CMAX=C
006087      000      CALL COPY(NPTS,COST,CHEMP)
006088      000      K1=J1
006089      000      K2=J2
006090      000      DO 35 J=1,NPTS
006091      000      35 ITEMP(J)=IPI(J)
006092      000      GO TO 25
006093      000      40 IT=IT+1
006094      000      IF(IT.LT.MAXIT)GO TO 40
006095      000      C*      GO BACK TO BEST COST
006096      000      CALL COPY(NPTS,CHEMP,COST)
006097      000      IF(IFOUT(JPI).NE.0)CALL IFOUT(JPI),1,MAXPTS-IT)
006098      000      WRITE(6,2001) CMAX
006099      000      CALL TIMEOUT
006100      000      IF(JPI.EQ.NPI)GO TO 800
006101      000      IT=0
006102      000      JPI=JPI+1
006103      000      PIPISAVE(JPI)
006104      000      WRITE(6,2100)PI
006105      000      J1=K1
006106      000      J2=K2
006107      000      DO 50 J=1,NPTS
006108      000      50 IPI(J)=ITEMP(J)
006109      000      GO TO 25
006110      000      800 FOUND=.TRUE.
006111      000      900 WRITE(6,2000) CSAVE,FOUND
006112      000      WRITE(6,2001) CMAX
006113      000      2001 FORMAT(' MAXIMUM COST',F10.3)
006114      000      CALL TIMEOUT
006115      000      RETURN
006116      000      END
006117      000      DELT,SI TUMER-S*FATHOM,TIMEOUT,,,199576103711
006118      000      SUBROUTINE TIMEOUT
006119      000      K=ITIME(1,J)
006120      000      T=1/5000.
006121      000      WRITE(6,2000)T
006122      000      2000 FORMAT(' TIME',F10.3)
006123      000      RETURN
006124      000      END
006125      000      DELT,SI TUMER-S*FATHOM,IRSTADNRV,,,199615101112
006126      000      FUNCTION TREE(N,IPI)
006127      000      C*      TREE FINDS THE COST OF A MINIMUM TREE
006128      000      INCLUDE BLANK
006129      000      DIMENSION IPI(1)
006130      000      DIMENSION CLINK(MAXPTS),LINK(MAXPTS)
006131      000      M=NPTS-N-1
006132      000      C=0.0
006133      000      IF(M.LE.0)GO TO 900
006134      000      C*      FIND MINIMUM COST
006135      000      C=1.0E38
006136      000      DO 20 I=1,NPTS
006137      000      IPI(I)=-2

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006036 000      IF(IN(I).OR.I.EQ.NPTS)GO TO 20
006039 000      L=L+1
006040 000      DO 10 J=L,NPTS
006041 000      IF(IN(J))GO TO 10
006042 000      IF(C.LE.COST(I,J))GO TO 20
006043 000      IROW=I
006044 000      ICOL=J
006045 000      C=COST(I,J)
006046 000      10 CONTINUE
006047 000      20 LINK(I)=0
006048 000      IPI(IROW)=1+IPI(IROW)
006049 000      IPI(ICOL)=1+IPI(ICOL)
006050 000      M=M-1
006051 000      IF(M.EQ.0)GO TO 900
006052 000      C*      FIND MINIMUM CONNECTION TO EACH POINT
006053 000      CMIN=1.0E38
006054 000      DO 50 I=1,NPTS
006055 000      IF(IN(I))GO TO 50
006056 000      IF(I.EQ.IROW.OR.I.EQ.ICOL)GO TO 50
006057 000      IF(COST(IROW,I).GT.COST(ICOL,I))GO TO 30
006058 000      LINK(I)=IROW
006059 000      CLINK(I)=COST(IROW,I)
006060 000      GO TO 40
006061 000      30 LINK(I)=ICOL
006062 000      CLINK(I)=COST(ICOL,I)
006063 000      40 IF(CLINK(I).GE.CMIN)GO TO 50
006064 000      JROW=LINK(I)
006065 000      JCOL=I
006066 000      CMIN=CLINK(I)
006067 000      50 CONTINUE
006068 000      IPI(JROW)=1+IPI(JROW)
006069 000      IPI(JCOL)=1+IPI(JCOL)
006070 000      ICOL=JCOL
006071 000      LINK(ICOL)=0
006072 000      C=C+CLINK(ICOL)
006073 000      M=M-1
006074 000      IF(M.EQ.0)GO TO 900
006075 000      C*      UPDATE MINIMUM CONNECTION
006076 000      DO 60 J=1,M
006077 000      CMIN=1.0E38
006078 000      DO 70 I=1,NPTS
006079 000      IF(LINK(I).EQ.0)GO TO 70
006080 000      IF(CLINK(I).LE.COST(I,ICOL))GO TO 60
006081 000      LINK(I)=ICOL
006082 000      CLINK(I)=COST(I,ICOL)
006083 000      60 IF(CLINK(I).GE.CMIN)GO TO 70
006084 000      JROW=LINK(I)
006085 000      JCOL=I
006086 000      CMIN=CLINK(I)
006087 000      70 CONTINUE
006088 000      IPI(JROW)=1+IPI(JROW)
006089 000      IPI(JCOL)=1+IPI(JCOL)
006090 000      ICOL=JCOL
006091 000      LINK(ICOL)=0
006092 000      80 C=C+CLINK(ICOL)
006093 000      900 TREE=C
006094 000      RETURN

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006695 000      END
006696 000  BELT,SI TUMER-S*FATHOM,IREF/OLD,.,176314121411
006697 000      FUNCTION TREE(N,IPI)
006698 000  C*      TREE FINDS THE COST OF A MINIMUM TREE
006699 000      INCLUDE BLANK
006700 000      DIMENSION IPI(1)
006701 000      DIMENSION LINK(MAXPTS)
006702 000      MNPTS=N-1
006703 000      NLINK=0
006704 000      C=0.0
006705 000      IF(M.LE.0)GO TO 900
006706 000      DO 10 I=1,NPTS
006707 000      IPI(I)=-2
006708 000  10 LINK(I)=0
006709 000      DO 60 I=1,NLIST
006710 000  C*      CHECK IF POINTS YH PHE
006711 000      J=LIST(1,I)
006712 000      K=LIST(2,I)
006713 000      IF((IN(J).OR.IN(K)).EQ.0)GO TO 30
006714 000  C*      CHECK IF POINTS ALREADY CONNECTED
006715 000      MX=MAX(LINK(J),LINK(K))
006716 000      IF(MX.EQ.0)GO TO 40
006717 000      IF(LINK(J).EQ.LINK(K))GO TO 40
006718 000  C*      ADD BRANCH TO TREE
006719 000      MN=MIN(LINK(J),LINK(K))
006720 000      IF(MN.EQ.0)GO TO 30
006721 000  C*      COMBINE LINKS
006722 000      DO 20 L=1,NPTS
006723 000      IF(LINK(L).EQ.MX)LINK(L)=MN
006724 000  20 CONTINUE
006725 000      GO TO 50
006726 000  C*      ADD TO EXISTING LINK
006727 000  30 LINK(J)=MX
006728 000      LINK(K)=MX
006729 000      GO TO 50
006730 000  C*      FORM NEW LINK
006731 000  40 NLINK=NLINK+1
006732 000      LINK(J)=NLINK
006733 000      LINK(K)=NLINK
006734 000  50 C=C+COST(J,K)
006735 000      IPI(J)=1+IPI(J)
006736 000      IPI(K)=1+IPI(K)
006737 000      M=M-1
006738 000      IF(M.EQ.0)GO TO 900
006739 000  60 CONTINUE
006740 000  900 TREE=C
006741 000      RETURN
006742 000      END
006743 000  BELT,SI TUMER-S*FATHOM,VOGEL,.,135013051012
006744 000      SUBROUTINE VOGELIN,COST,PRINTV,Z,ISAVE)
006745 000      PARAMETER MAXPTS=50
006746 000      DIMENSION COST(MAXPTS,MAXPTS),ISAVE(MAXPTS)
006747 000      DIMENSION LROW(MAXPTS),LCOL(MAXPTS),ITOP(MAXPTS),NODE(MAXPTS)
006748 000      LOGICAL PRINTV
006749 000      CMAX=1.0E38
006750 000  C*** INITIALIZE VARIABLES
006751 000      ICOUNT=1

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006752 000 ITREE=0
006753 000 DO 15 I=1,N
006754 000 LROW(I)=0
006755 000 LCOL(I)=0
006756 000 NODE(I)=0
006757 000 15 CONTINUE
006758 000 Z=0.
006759 000 16 CONTINUE
006760 000 ROWPEN=0.
006761 000 COLPEN=0.
006762 000 C*** FIGURE ROW PENALTIES
006763 000 DO 50 I=1,N
006764 000 IF (LROW(I).EQ.1) GO TO 50
006765 000 CMIN=CMAX
006766 000 DO 20 J=1,N
006767 000 IF (I.EQ.J) GO TO 20
006768 000 IF (ICOUNT.NE.N.AND.NODE(I).EQ.NODE(J).AND.NODE(I).NE.0) GO TO 20
006769 000 IF (LCOL(J).EQ.1.OR.COST(I,J).GT.CMIN) GO TO 20
006770 000 MINI=J
006771 000 CMIN=COST(I,J)
006772 000 20 CONTINUE
006773 000 DIFROW=CMAX
006774 000 DIFF=CMAX
006775 000 DO 25 J=1,N
006776 000 IF (I.EQ.J) GO TO 25
006777 000 IF (ICOUNT.NE.N.AND.NODE(I).EQ.NODE(J).AND.NODE(I).NE.0) GO TO 25
006778 000 IF (LCOL(J).EQ.1.OR.COST(I,J).GT.DIFF) GO TO 25
006779 000 DIFF=1ABS(COST(I,J)-DIFF)
006780 000 IF (DIFROW.GT.DIFF) DIFROW=DIFF
006781 000 25 CONTINUE
006782 000 IF (ROWPEN.GE.DIFROW) GO TO 50
006783 000 ROWPEN=DIFROW
006784 000 IROW=I
006785 000 JROW=MINI
006786 000 50 CONTINUE
006787 000 C*** FIGURE COLUMN PENALTIES
006788 000 DO 100 J=1,N
006789 000 IF (LCOL(J).EQ.1) GO TO 100
006790 000 CMIN=CMAX
006791 000 DO 60 I=1,N
006792 000 IF (I.EQ.J) GO TO 60
006793 000 IF (ICOUNT.NE.N.AND.NODE(I).EQ.NODE(J).AND.NODE(J).NE.0) GO TO 60
006794 000 IF (LROW(I).EQ.1.OR.COST(I,J).GT.CMIN) GO TO 60
006795 000 MINI=I
006796 000 CMIN=COST(I,J)
006797 000 60 CONTINUE
006798 000 DIFCOL=CMAX
006799 000 DIFF=CMAX
006800 000 DO 65 I=1,N
006801 000 IF (I.EQ.J) GO TO 65
006802 000 IF (ICOUNT.NE.N.AND.NODE(I).EQ.NODE(J).AND.NODE(J).NE.0) GO TO 65
006803 000 IF (LROW(I).EQ.1.OR.COST(I,J).GT.DIFF) GO TO 65
006804 000 DIFF=ABS(COST(I,J)-DIFF)
006805 000 IF (DIFCOL.GT.DIFF) DIFCOL=DIFF
006806 000 65 CONTINUE
006807 000 IF (COLPEN.GE.DIFCOL) GO TO 100
006808 000 COLPEN=DIFCOL

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006809      000      ICOL=MINI
006810      000      JCOL=J
006811      000      100 CONTINUE
006812      000      C*** WE NOW HAVE ROW AND COLUMN DUALITIES.
006813      000      C*** DELETE THE ROW & THE COLUMN WITH THE MAX FLUX.
006814      000      IF (COLPEN.GT.ROWPEN) GO TO 105
006815      000      Z=Z+COST(IROW,JROW)
006816      000      LROW(IROW)=1
006817      000      LCOL(JROW)=1
006818      000      COST(JROW,IROW)=CHMAX
006819      000      IFOUR(IROW)=JROW
006820      000      IF (PRINTV) WRITE(6,1005) (COUNT,IROW,JROW)
006821      000      IF (NODE(IROW).NE.0.OR.NODE(JROW).NE.0) GO TO 105
006822      000      ITRCE=ITREE+1
006823      000      NODE(IROW)=ITREE
006824      000      NODE(JROW)=ITREE
006825      000      GO TO 115
006826      000      105 CONTINUE
006827      000      IF (NODE(IROW).EQ.0.OR.NODE(JROW).EQ.0) GO TO 106
006828      000      ITOP=MAX(NODE(IROW),NODE(JROW))
006829      000      IREP=MIN(NODE(IROW),NODE(JROW))
006830      000      GO TO 113
006831      000      106 CONTINUE
006832      000      IF (NODE(IROW).EQ.0) NODE(IROW)=NODE(JROW)
006833      000      NODE(JROW)=NODE(IROW)
006834      000      GO TO 115
006835      000      110 Z=Z+COST(ICOL,JCOL)
006836      000      LROW(ICOL)=1
006837      000      LCOL(JCOL)=1
006838      000      COST(JCOL,ICOL)=CHMAX
006839      000      ITOUR(ICOL)=JCOL
006840      000      IF (PRINTV) WRITE(6,1006) (COUNT,ICOL,JCOL)
006841      000      IF (NODE(ICOL).NE.0.OR.NODE(JCOL).NE.0) GO TO 111
006842      000      ITRCE=ITREE+1
006843      000      NODE(ICOL)=ITREE
006844      000      NODE(JCOL)=ITREE
006845      000      GO TO 115
006846      000      111 CONTINUE
006847      000      IF (NODE(ICOL).EQ.0.OR.NODE(JCOL).EQ.0) GO TO 112
006848      000      ITOP=MAX(NODE(ICOL),NODE(JCOL))
006849      000      IREP=MIN(NODE(ICOL),NODE(JCOL))
006850      000      GO TO 113
006851      000      112 CONTINUE
006852      000      IF (NODE(ICOL).EQ.0) NODE(ICOL)=NODE(JCOL)
006853      000      NODE(JCOL)=NODE(ICOL)
006854      000      GO TO 115
006855      000      113 DO 114 I=1,N
006856      000          IF (NODE(I).EQ.ITOP) NODE(I)=IREP
006857      000      114 CONTINUE
006858      000      115 IF (ICOUNT.EQ.N) GO TO 1020
006859      000      1020 FORMAT(1X,I3,1H., ' CITY ',I3, ' TO CITY ',I3)
006860      000      ICOUNT=ICOUNT+1
006861      000      C      DO 130 I=1,N
006862      000      C      IF (PRINTV) WRITE(6,1015) (COST(I,J),J=1,N)
006863      000      C 130 CONTINUE
006864      000      GO TO 16
006865      000      999 CONTINUE

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006860 000 I=ISAVE(1)
006867 000 DO 200 J=2,N
006860 000 I=ITOUR(I)
006869 000 200 ISAVE(J)=I
006870 000 CALL OUTPUT(N,Z,ISAVE)
006871 000 C*** RESTORE MATRIX
006872 000 DO 300 I=1,N
006873 000 J=ITOUR(I)
006874 000 300 COST(J,I)=COST(I,J)
006875 000 RETURN
006876 000 END
006877 000 @PDP,FI TUMER-S*FATHOM.COMMON/BLANK,,,150624070312
006878 000 BLANK PROC
006879 000 PARAMETER MAXPTS=60,MXLIST=1160
006880 000 COMMON NPTS,NLIST,COSVE,EX(MAXPTS),LIST(2,MXLIST),
006881 000 * ISAVE(MAXPTS),COST(MAXPTS,MAXPTS),EX(MAXPTS,MAXPTS)
006882 000 *,FOUND
006883 000 LOGICAL IN,EX,FOUND
006884 000 END
006885 000 @ELT,SI TUMER-S*NONSYM.CITY,,,144433070512
006886 000 @ASG,A TUMER-S*NONSYM.
006887 000 @ASG,A TUMER-S*MATRIX.
006888 000 @USE 9,MATRIX.
006889 000 @ADD NONSYM.MAP-XOT
006890 000 9,0,1.0E30,0,,20,3,0,0,0,0,0
006891 000 0,0,9
006892 000 5,,1,,0.2
006893 000 @ADD FATHOM-DATA.CITY/9
006894 000 92.
006895 000 95.
006896 000 97.
006897 000 100.
006898 000 @EOF
006899 000 @ELT,SI TUMER-S*NONSYM.CMINP3,,,120154063012
006900 000 FUNCTION CMINP3(N,J1,J2)
006901 000 C* FIND COST OF BEST CONNECTION
006902 000 C* FROM N TO THE ONE TREE
006903 000 C* J1 FROM N , J2 TO N
006904 000 INCLUDE BLANK
006905 000 J2=MINPT(N,1)
006906 000 IF(J2.NE.0) GO TO 5
006907 000 C01=1.0E38
006908 000 GO TO 40
006909 000 5 C=1.0E38
006910 000 DO 10 J=1,NPTS
006911 000 IF(J.EQ.N.OR.J.EQ.J2) GO TO 10
006912 000 IF(C.LE.COST(J,N)) GO TO 10
006913 000 C=COST(J,N)
006914 000 J1=J
006915 000 10 CONTINUE
006916 000 C01=COST(N,J2)+COST(J1,N)
006917 000 C=1.0E38
006918 000 DO 20 J=1,NPTS
006919 000 IF(J.EQ.N) GO TO 20
006920 000 IF(C.LE.COST(J,N)) GO TO 20
006921 000 C=COST(J,N)
006922 000 J4=J

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006923 000 20 CONTINUE
006924 000 J3=0
006925 000 C=1.0E30
006926 000 DO 30 J=1,NPTS
006927 000 IF(J.EQ.N.OR.J.EQ.01.0012(1,J)) GO TO 30
006928 000 IF(C.LE.COST(N,J)) GO TO 30
006929 000 C=COST(N,J)
006930 000 J3=J
006931 000 30 CONTINUE
006932 000 IF(J3.EQ.0) GO TO 40
006933 000 C02=COST(N,J3)+COST(J3,11)
006934 000 IF(C01.LE.C02) GO TO 40
006935 000 C01=C02
006936 000 J1=J4
006937 000 J2=J3
006938 000 40 CMINP3=C01
006939 000 RETURN
006940 000 END
006941 000 DELT,SI TUMER-S*NONSYM,COPY,171750051012
006942 000 SUBROUTINE COPY(N,NPT)
006943 000 C* COPY COPIES ONE ARRAY INTO ANOTHER
006944 000 PARAMETER MAXPTS=100
006945 000 DIMENSION A(MAXPTS,MAXPTS),B(MAXPTS,MAXPTS)
006946 000 DO 10 I=1,N
006947 000 DO 10 J=1,N
006948 000 10 B(I,J)=A(I,J)
006949 000 RETURN
006950 000 END
006951 000 DELT,SI TUMER-S*NONSYM.FOR,171750063412
006952 000 @FOR,Y NONSYM.TIMEOUT
006953 000 @FOR,Y NONSYM.IO
006954 000 @FOR,Y NONSYM.COPY
006955 000 @FOR,Y NONSYM.MINPT
006956 000 @FOR,Y NONSYM.SHOW/BATCH,SHOW
006957 000 @FOR,Y NONSYM.INPUT
006958 000 @FOR,Y NONSYM.MAIN
006959 000 @FOR,Y NONSYM.OUTPUT/BATCH,OUTPUT
006960 000 @FOR,Y NONSYM.START
006961 000 @FOR,Y NONSYM.CMINP3
006962 000 @FOR,Y NONSYM.TREE
006963 000 @FOR,Y NONSYM.IMRAND
006964 000 @FOR,Y NONSYM.MINPT2
006965 000 @EOF
006966 000 DELT,SI TUMER-S*NONSYM.INPUT,171750063312
006967 000 SUBROUTINE INPUT(NPTS,MAXPTS,COST,PRINTM)
006968 000 C* INPUT READS NONSYM ORIC COST MATRIX
006969 000 PARAMETER MAXPTS=100
006970 000 DIMENSION COST(MAXPTS,MAXPTS)
006971 000 LOGICAL PRINTM
006972 000 IF(IFIN.NE.0)GO TO 20
006973 000 C* READ MATRIX
006974 000 DO 10 I=1,NPTS
006975 000 10 READ(5,1000) (COST(I,J),J=1,NPTS)
006976 000 1000 FORMAT()
006977 000 GO TO 30
006978 000 C* USE MATRIX PREVIOUSLY SAVED
006979 000 20 CALL IO(IFIN,2,NPTS,COST)

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006980 000 30 IF (PRINTM) CALL SHOW (NPTS,COST)
006981 000 CALL TIMEOUT
006982 000 RETURN
006983 000 END
006984 000 DELT,SI TUMER-S*NONSYM.10,,,1017001063212
006985 000 SUBROUTINE INRAND (NPTS,IOPT,PRINTM)
006986 000 C* INRAND GENERATES A LOWER TRIANGLE COST MATRIX RANDOMLY
006987 000 C* A=LOWER LIMIT, B=UPPER LIMIT
006988 000 PARAMETER MAXPTS=80
006989 000 DIMENSION COST (MAXPTS,MAXPTS),X (6325)
006990 000 LOGICAL PRINTM
006991 000 IF (IFIN.NE.0) GO TO 20
006992 000 C* GENERATE COST MATRIX
006993 000 READ (5,1000) A,B,X (1)
006994 000 1000 FORMAT ( )
006995 000 NUM=(NPTS*NPTS)-NPTS+1
006996 000 CALL RANDU (X,NUM)
006997 000 IJ=1
006998 000 DO 10 I=1,NPTS
006999 000 COST (I,I)=1.0E38
007000 000 DO 10 J=1,NPTS
007001 000 IF (I.EQ.J) GO TO 10
007002 000 IJ=IJ+1
007003 000 IYFL=A+(B-A)*X (IJ)
007004 000 COST (I,J)=IYFL
007005 000 10 CONTINUE
007006 000 GO TO 30
007007 000 20 CALL IO (IFIN,2,NPTS,COST)
007008 000 30 IF (PRINTM) CALL SHOW (NPTS,COST)
007009 000 CALL TIMEOUT
007010 000 RETURN
007011 000 END
007012 000 DELT,SI TUMER-S*NONSYM.10,,,1017001012
007013 000 SUBROUTINE IO (IFILE,IOPT,N)
007014 000 C* IO READS OR WRITES A MATRIX IN A FILE
007015 000 C* IOPT=1 WRITE
007016 000 C* IOPT=2 READ
007017 000 PARAMETER MAXPTS=80
007018 000 DIMENSION A (MAXPTS,MAXPTS)
007019 000 LOGICAL BAD
007020 000 BAD=.FALSE.
007021 000 5 CALL NTRAN (IFILE,10,22)
007022 000 DO 10 I=1,N
007023 000 CALL NTRAN (IFILE,IOPT,N,A (1,I),ISTAT,22)
007024 000 IF (ISTAT.LE.0) GO TO 700
007025 000 10 CONTINUE
007026 000 RETURN
007027 000 700 IF (BAD) GO TO 800
007028 000 BAD=.TRUE.
007029 000 GO TO 5
007030 000 800 WRITE (6,2000) IFILE
007031 000 2000 FORMAT (' FILE ERROR ON UNIT',I3)
007032 000 STOP
007033 000 END
007034 000 DELT,SI TUMER-S*NONSYM.MAIN,,,102150100512
007035 000 C* THIS PROGRAM SOLVES THE TRAVELING SALESMAN PROBLEM
007036 000 C* BY FATHOMING

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007037 000      INCLUDE BLANK
007038 000      DIMENSION PISAVE(20), IFOUT(20)
007039 000      DIMENSION IPOINT(100), IPOINT(100), ITEMP(100), ITEMPO(100), MAXPTS(100)
007040 000      DIMENSION ACOST(100), ACPCT(100), EPS(100), MAXIT(100), NPTS(100)
007041 000      DIMENSION ASAVE(100), ACPCT(100), EPS(100), MAXIT(100), NPTS(100)
007042 000      LOGICAL EXSAVE, PRINTL, PRINTM, PRINTV, ALR
007043 000  C*      READ DATA
007044 000      READ(5,1000)NPTS,ISTART,CSAVE,EPS,MAXIT,NPTS,IFIN,L,M,K,MATGEN
007045 000      WRITE(6,1001) NPTS,ISTART,MAXIT,NPTS
007046 000 1001  FORMAT(//2X,'NPTS=',I5,' ISTART=',I4,' EPS=',F6.2,
007047 000      *' MAXIT=',I5,' NPTS=',I5)
007048 000 1002  FORMAT(//2X,'PI VALUE=',F6.1)
007049 000 1000  FORMAT(//
007050 000      IF(NPTS.EQ.0)GO TO 2
007051 000      READ(5,1000) (IPOINT(I),I=1,NPTS)
007052 000      READ(5,1000) (IPOINT(I),I=1,NPTS)
007053 000  C*      NPTS=NUMBER OF POINTS
007054 000  C*      ISTART=STARTING POINT NUMBER IF ISTART=0
007055 000  C*      CSAVE=INITIAL COST OF CONNECTION
007056 000  C*      EPS=STOPPING CRITERION DIFFERENCE BETWEEN BOUNDS
007057 000  C*      MAXIT=MAXIMUM NUMBER OF ITERATIONS
007058 000  C*      NPTS=NUMBER OF POINTS
007059 000  C*      IFIN=INPUT FILE FOR COST MATRIX
007060 000  C*      IFOUT=OUTPUT FILE FOR COST MATRIX
007061 000  C*      PISAVE=VALUES FOR COST MATRIX
007062 000  C*      PRINTL=PRINT OPTION FOR COST MATRIX
007063 000  C*      PRINTM=PRINT OPTION FOR COST MATRIX
007064 000  C*      PRINTV=PRINT OPTION FOR COST MATRIX
007065 000  C*      MATGEN=COST MATRIX IS GENERATED IN RANDOM MANNER(=1)
007066 000 2  PRINTL=(L.GT.0)
007067 000      PRINTM=(M.GT.0)
007068 000      PRINTV=(K.GT.0)
007069 000      IF(MATGEN.EQ.1) GO TO 3
007070 000      CALL INPUT(NPTS,IFIN,PRINTM,PRINTV)
007071 000      GO TO 4
007072 000 3  CALL INRAND(NPTS,IFIN,PRINTM,PRINTV)
007073 000  C*      FIND STARTING POINT AND ADJUST COST MATRIX
007074 000 4  FOUND=.FALSE.
007075 000      CALL START(ISTART,MAXIT,NPTS,PISAVE,IFOUT,PRINTM,PRINTV)
007076 000      IF(FOUND) GO TO 5
007077 000  C*      READ COST FACTOR FOR PATHFINDING
007078 000      READ(5,1000,END=2001)CFACT
007079 000      CFACT=CPCT/100.0
007080 000      CMIN=CMAX
007081 000      IF(CPCT.LE.0.0)CMAX=CMAX
007082 000      IF(CPCT.GT.0.0)CMAX=CMAX
007083 000      FOUND=.FALSE.
007084 000  C*      PLACE STARTING POINT ON LIST
007085 000 5  NPTSIN=1
007086 000      IPOINT(1)=ISTART
007087 000      LAST=ISTART
007088 000  C*      FIND MINIMUM CONNECTION
007089 000      C=CMINP3(ISTART,J1,J2)
007090 000  C*      FIND MINIMUM TREE
007091 000      TCOST(1)=TREE(1,ITEMP,ITEMPO)
007092 000  C*      COMPUTE COST
007093 000      C=C+TCOST(1)

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007094 000 IF(CMIN.LT.C)CMIN=C
007095 000 IF(PRINTL)CALL OUTPUT(1,C,IPOINT)
007096 000 C* ADD NEW POINT
007097 000 10 NPTSIN=NPTSIN+1
007098 000 IF(NPTSIN.EQ.NPTS)GO TO 50
007099 000 DO 20 I=1,NPTS
007100 000 20 EX(NPTSIN,I)=.FALSE.
007101 000 ACOST(NPTSIN)=ACOST(NPTSIN-1)+COST(LAST,J2)
007102 000 LAST=J2
007103 000 IPOINT(NPTSIN)=LAST
007104 000 IN(LAST)=.TRUE.
007105 000 C* FIND MINIMUM CONNECTIONS
007106 000 25 IF(NPTSIN.NE.1) GO TO 30
007107 000 C=CMINP3(ISTART,J1,J2)
007108 000 GO TO 35
007109 000 30 J1=MINPT2(ISTART)
007110 000 J2=MINPT(LAST,NPTSIN)
007111 000 35 IF(J2.EQ.0)GO TO 40
007112 000 C* FIND MINIMUM TREE
007113 000 TCOST(NPTSIN)=TREL(MINPT(LAST,NPTSIN),TEMP0)
007114 000 C* COMPUTE COST
007115 000 C=ACOST(NPTSIN)+TCOST(MINPT(LAST,NPTSIN),J1,ISTART)+COST(LAST,J2)
007116 000 IF(PRINTL)CALL OUTPUT(NPTSIN,C,IPOINT)
007117 000 IF(C.LE.CMAX)GO TO 10
007118 000 C* REMOVE POINT FROM LIST
007119 000 40 NPTSIN=NPTSIN-1
007120 000 IF(NPTSIN.EQ.0)GO TO 80
007121 000 EX(NPTSIN,LAST)=.TRUE.
007122 000 IN(LAST)=.FALSE.
007123 000 LAST=IPOINT(NPTSIN)
007124 000 GO TO 25
007125 000 C* COMPUTE COST OF COMPLETE TREE
007126 000 50 C=ACOST(NPTS-1)+COST(J1,ISTART)+COST(LAST,J2)
007127 000 LAST=J2
007128 000 IPOINT(NPTS)=LAST
007129 000 IF(CMAX.LT.C)GO TO 40
007130 000 WRITE(6,2000)
007131 000 2000 FORMAT(//,' NEW SOLUTION')
007132 000 CALL OUTPUT(NPTS,C,IPOINT)
007133 000 CALL TIMEOUT
007134 000 IF(CSAVE.LE.C.AND.FOUND)GO TO 40
007135 000 CSAVE=C
007136 000 IF(CPCT.GT.0.0)GO TO 52
007137 000 C* ADJUST UPPER BOUND
007138 000 CMAX=(C+CMIN)/2.0
007139 000 IF((CMAX-CMIN).LE.EPS)GO TO 90
007140 000 GO TO 55
007141 000 52 CMAX=CFAC*CSAVE
007142 000 55 FOUND=.TRUE.
007143 000 DO 60 I=1,NPTS
007144 000 ISAVE(I)=IPOINT(I)
007145 000 ASAVE(I)=ACOST(I)
007146 000 TSAVE(I)=TCOST(I)
007147 000 DO 60 J=1,NPTS
007148 000 60 EXSAVE(I,J)=EX(I,J)
007149 000 IF(CMAX.LT.CMIN) GO TO 80
007150 000 GO TO 40

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007151 000 C* ADJUST LOWER BOUND
007152 000 80 IF(CPCT.GT.0.0)GO TO 100
007153 000 CMIN=CMAX
007154 000 CMAX=(CMIN+CSAVE)/2.0
007155 000 IF((CMAX-CMIN).LT.1.E-10)GO TO 90
007156 000 GO TO 105
007157 000 90 GO TO 101
007158 000 C* PRINT FINAL SOLUTION
007159 000 100 CONTINUE
007160 000 WRITE(6,2100)CPCT
007161 000 2100 FORMAT(//,1X,F5.1,' % SOLUTION')
007162 000 CALL TIMEOUT
007163 000 IF(.NOT.ALR) GO TO 101
007164 000 FOUND=.FALSE.
007165 000 ALR=.FALSE.
007166 000 CSAVE=1.0F38
007167 000 CPCT=0.
007168 000 C* READ COST FACTOR FROM KEYRING
007169 000 101 READ(5,1000,END=900)CF
007170 000 IF(CPCT1.LE.CPCT) GO TO 101
007171 000 CPCT=CPCT1
007172 000 102 CFACT=CPCT/100.0
007173 000 CMAX=CFACT+CSAVE
007174 000 105 IF(.NOT.FOUND)GO TO 100
007175 000 DO 110 I=1,NPTS
007176 000 IN(I)=.TRUE.
007177 000 IPOINT(I)=ISAVE(I)
007178 000 ACOST(I)=ASAVE(I)
007179 000 TCOST(I)=TSAVE(I)
007180 000 DO 110 J=1,NPTS
007181 000 110 EX(I,J)=EXSAVE(I,J)
007182 000 LAST=IPOINT(NPTS)
007183 000 NPTSIN=NPTS
007184 000 GO TO 40
007185 000 120 CONTINUE
007186 000 DO 130 I=1,NPTS
007187 000 DO 130 J=1,NPTS
007188 000 130 EX(I,J)=.FALSE.
007189 000 GO TO 5
007190 000 800 WRITE(6,3000)
007191 000 3000 FORMAT(//,' SOLVED ON ASCENT')
007192 000 CPCT=100.0
007193 000 CMIN=CMAX
007194 000 FOUND=.FALSE.
007195 000 GO TO 102
007196 000 900 CALL TIMEOUT
007197 000 STOP
007198 000 END
007199 000 DELT,SI TUMER-S*NONSYM,MAPS-XOT,,,134275062312
007200 000 @MAP,S
007201 000 IN NONSYM.
007202 000 LIB SYSTEMS*MATHSTAT.
007203 000 END
007204 000 @XQT
007205 000 DELT,SI TUMER-S*NONSYM,MAP-XOT,,,125316062112
007206 000 @MAP,N
007207 000 IN NONSYM.

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007200 000 LIB SYSTEM=MATHSTAT.
007209 000 END
007210 000 MAXOT
007211 000 DELT,SI TUMER-S*NONSYS.MINPT,.,,133700113111
007212 000 FUNCTION MINPT(I,N)
007213 000 C* MINPT FINDS POINT WITH MINIMUM COST FOR CONNECTION TO
007214 000 C* GIVEN POINT
007215 000 INCLUDE BLANK
007216 000 M=0
007217 000 C=1.0E38
007218 000 DO 10 J=1,NPTS
007219 000 IF(IN(J).OR.EX(N,J))GO TO 10
007220 000 IF(C.LE.COST(I,J))GO TO 10
007221 000 C=COST(I,J)
007222 000 M=J
007223 000 10 CONTINUE
007224 000 MINPT=M
007225 000 RETURN
007226 000 END
007227 000 DELT,SI TUMER-S*NONSYS.MINPT2,.,,133700113112
007228 000 FUNCTION MINPT2(I)
007229 000 C* MINPT2 FINDS POINT WITH LOWEST COST FOR CONNECTION TO
007230 000 C* GIVEN POINT
007231 000 C* I=GIVEN POINT
007232 000 INCLUDE BLANK
007233 000 M=0
007234 000 C=1.0E38
007235 000 DO 10 J=1,NPTS
007236 000 IF(J.EQ.I.OR.IN(J)) GO TO 10
007237 000 IF(C.LE.COST(J,I)) GO TO 10
007238 000 C=COST(J,I)
007239 000 M=J
007240 000 10 CONTINUE
007241 000 MINPT2=M
007242 000 RETURN
007243 000 END
007244 000 DELT,SI TUMER-S*NONSYS.ORDER/ORD,.,,170251121411
007245 000 SUBROUTINE ORDER(ISTART)
007246 000 C* ORDER ORDERS LIST OF COMBINATIONS BY INCREASING COST
007247 000 C* COMBINATIONS CONTAINING STARTING POINT ARE OMITTED
007248 000 INCLUDE BLANK
007249 000 N=NPTS-1
007250 000 NLIST=(N*N-N)/2
007251 000 IF(ISTART.EQ.0)NLIST=(N*NPTS)/2
007252 000 DO 30 K=1,NLIST
007253 000 C=1.0E38
007254 000 DO 20 I=1,N
007255 000 IF(I.EQ.ISTART)GO TO 20
007256 000 L=I+1
007257 000 DO 10 J=L,NPTS
007258 000 IF(J.EQ.ISTART.OR.EX(I,J))GO TO 10
007259 000 IF(C.LE.COST(I,J))GO TO 10
007260 000 C=COST(I,J)
007261 000 LIST(1,K)=I
007262 000 LIST(2,K)=J
007263 000 10 CONTINUE
007264 000 20 CONTINUE

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007265 000 I=LIST(1,K)
007266 000 J=LIST(2,K)
007267 000 30 EX(I,J)=.TRUE.
007268 000 DO 40 I=1,NPTS
007269 000 DO 40 J=1,NPTS
007270 000 40 EX(I,J)=.FALSE.
007271 000 RETURN
007272 000 END
007273 000 DELT,SI TUMER-S*NONSYM.OUTPUT/PATH,,,110552062912
007274 000 SUBROUTINE OUTPUT(NC,IPOINT)
007275 000 PARAMETER MAXPTS=50
007276 000 C* OUTPUT PRINTS COST AND LIST OF POINTS
007277 000 DIMENSION IPOINT(MAXPTS)
007278 000 M=MIN(N,25)
007279 000 WRITE(6,2000)C,(IPOINT(L),L=1,M)
007280 000 IF(N.LE.25)GO TO 900
007281 000 DO 10 L=26,N,25
007282 000 M=MIN(N,L+24)
007283 000 10 WRITE(6,2100) (IPOINT(L),L=1,M)
007284 000 2000 FORMAT(' COST',F7.0,4X,'IPOINT',25I4)
007285 000 2100 FORMAT(22X,25I4)
007286 000 900 RETURN
007287 000 END
007288 000 DELT,SI TUMER-S*NONSYM.OUTPUT/PATH,,,172614113311
007289 000 SUBROUTINE OUTPUT(NC,IPOINT)
007290 000 C* OUTPUT PRINTS COST AND LIST OF POINTS
007291 000 DIMENSION IPOINT(1)
007292 000 M=MIN(N,12)
007293 000 WRITE(6,2000)C,(IPOINT(L),L=1,M)
007294 000 IF(N.LE.12)GO TO 900
007295 000 DO 10 L=13,N,12
007296 000 M=MIN(N,L+11)
007297 000 10 WRITE(6,2100) (IPOINT(L),L=1,M)
007298 000 2000 FORMAT(' COST',F7.0,4X,'IPOINT',12I4)
007299 000 2100 FORMAT(22X,12I4)
007300 000 900 RETURN
007301 000 END
007302 000 DELT,SI TUMER-S*NONSYM.RANDU,,,160744073712
007303 000 BASG,A TUMER-S*NONSYM.
007304 000 BASG,A TUMER-S*MATRIX.
007305 000 BUSE 9,MATRIX.
007306 000 BADD NONSYM.MAP-XGT
007307 000 30,0,1.0E30,15.,30,3,0,0,0,0,1
007308 000 0,0,9
007309 000 5.,1.,.2
007310 000 100.,3500.,1.111
007311 000 0.
007312 000 92.
007313 000 95.
007314 000 97.
007315 000 99.
007316 000 100.
007317 000 BEOF
007318 000 DELT,SI TUMER-S*NONSYM.READ/IN,,,101664070312
007319 000 BASG,A TUMER-S*NONSYM.
007320 000 BASG,A FATHOM-DATA.
007321 000 BASG,A TUMER-S*MATRIX.

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007322 000 QUSE 9,MATRIX.
007323 000 QADD NONSYM,MAP=XQT
007324 000 6,0,1.0E30,0.,30,3,0,0,0,0,0
007325 000 0,0,9
007326 000 5.,1.,0.2
007327 000 QADD FATHOM-DATA.NONSYM6
007328 000 95.
007329 000 97.
007330 000 99.9
007331 000 100.
007332 000 QEOF
007333 000 QFLT,SI TUMER-S*NONSYM,RFOR,,,131017063312
007334 000 QRFUR,CS NONSYM,TIMOUT
007335 000 QRFUR,CS NONSYM,IO
007336 000 QRFUR,CS NONSYM,COPY
007337 000 QRFUR,CS NONSYM,MINPT
007338 000 QRFUR,CS NONSYM.SHOW/BATCH,SHOW
007339 000 QRFUR,SC FATHOM.INPUT,NONSYM.INPUT
007340 000 QRFUR,CS NONSYM.MAIN
007341 000 QRFUR,CS NONSYM.OUTPUT/BATCH,OUTPUT
007342 000 QRFUR,CS NONSYM.START
007343 000 QRFUR,CS NONSYM.CMINP3
007344 000 QRFUR,CS NONSYM.TRUE
007345 000 QRFUR,CS NONSYM.INRAND
007346 000 QRFUR,CS NONSYM.MINPT2
007347 000 QEOF
007348 000 QFLT,SI TUMER-S*NONSYM,SHOW/BATCH,,,134704051012
007349 000 SUBROUTINE SHOW(NPTS,COST)
007350 000 C* SHOW PRINTS COST MATRIX.
007351 000 PARAMETER MAXPTS=60
007352 000 DIMENSION COST(MAXPTS,MAXPTS)
007353 000 WRITE(6,2000)
007354 000 2000 FORMAT(' COST MATRIX')
007355 000 DO 20 K=1,NPTS,16
007356 000 L=MIN(NPTS,K+15)
007357 000 WRITE(6,2100) (J,J=K,L)
007358 000 DO 20 I=1,NPTS
007359 000 20 WRITE(6,2200) I,(COST(I,J),J=K,L)
007360 000 2100 FORMAT(4X,16I7)
007361 000 2200 FORMAT(1X,I3,16F7.0)
007362 000 RETURN
007363 000 END
007364 000 QFLT,SI TUMER-S*NONSYM,SHOW/DEMAND,,,135036051012
007365 000 SUBROUTINE SHOW(NPTS,COST)
007366 000 C* SHOW PRINTS COST MATRIX
007367 000 PARAMETER MAXPTS=60
007368 000 DIMENSION COST(MAXPTS,MAXPTS)
007369 000 WRITE(6,2000)
007370 000 2000 FORMAT(' COST MATRIX')
007371 000 DO 20 K=1,NPTS,9
007372 000 L=MIN(NPTS,K+8)
007373 000 WRITE(6,2100) (J,J=K,L)
007374 000 DO 20 I=1,NPTS
007375 000 20 WRITE(6,2200) I,(COST(I,J),J=K,L)
007376 000 2100 FORMAT(4X,9I7)
007377 000 2200 FORMAT(1X,I3,9F7.0)
007378 000 RETURN

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007379 000      END
007380 000      DELT,SI TUMER-S*NONSYM,START,1,1,1,0,0312
007381 000      SUBROUTINE START(IPI,IP0,PI,SAVE,IPI0,PRINTM,PRINTV)
007382 000      C*      START FINDS THE STARTING POINT AND ADJUSTS THE COST MATRIX
007383 000      INCLUDE BLANK
007384 000      DIMENSION PISAVE(100),IPI(100),IP0(100),PRINTM(100),PRINTV(100)
007385 000      DIMENSION IPI0(100),IP0(100),IPI(100),IP0(100),IPI(100),IP0(100)
007386 000      *IPI0(MAXPTS),IPI0(MAXPTS),IPI0(MAXPTS),IPI0(MAXPTS),IPI0(MAXPTS),IPI0(MAXPTS)
007387 000      LOGICAL PRINTM,PRINTV
007388 000      C*      ORDER CONNECTIONS BY INCREASING COST
007389 000      DO 5 I=1,NPTS
007390 000      5  IN(I)=.FALSE.
007391 000      IF(ISTART.NE.0) GO TO 20
007392 000      CMAX=-1.0030
007393 000      DO 20 I=1,NPTS
007394 000      IN(I)=.TRUE.
007395 000      T=TREE(1,ITEMP,I,IP0)
007396 000      C=CHIMP3(I,K1,K2)
007397 000      C=C+T
007398 000      IF(C.LE.CMAX) GO TO 20
007399 000      ISTART=I
007400 000      CMAX=C
007401 000      20  IN(I)=.FALSE.
007402 000      22  IN(ISTART)=.TRUE.
007403 000      T=TREE(1,ITEMP,ITEMP)
007404 000      C=CHIMP3(ISTART,K1,K2)
007405 000      CMAX=C+T
007406 000      WRITE(6,2000) CMAX,ISTART
007407 000      2000 FORMAT(' COST',F10.4,' END POINT',I3)
007408 000      ISAVE(1)=ISTART
007409 000      C*      USE VOGEL METHOD TO FIND SOLUTION
007410 000      IF(NPI.LE.0)GO TO 300
007411 000      CALL COPY(NPTS,COST,CIPI)
007412 000      CALL TIMEOUT
007413 000      C*      ITERATE ON COST MATRIX TO FIND MAXIMUM LOWER BOUND
007414 000      JPI=1
007415 000      PI=PISAVE(1)
007416 000      WRITE(6,2100)PI
007417 000      2100 FORMAT(' PI=',F10.4)
007418 000      IF=0
007419 000      23  J1=K1
007420 000      J2=K2
007421 000      DO 24 J=1,NPTS
007422 000      IPO(J)=ITEMP0(J)
007423 000      24  IPI(J)=ITEMP(J)
007424 000      C*      ADJUST COST
007425 000      25  IPI(ISTART)=0
007426 000      IPO(ISTART)=0
007427 000      IPI(J2)=IPI(J2)+1
007428 000      IPO(J1)=IPO(J1)+1
007429 000      ICOST=0
007430 000      DO 30 J=1,NPTS
007431 000      ICOST=ICOST+ABS(IPI(J))-ABS(IPO(J))
007432 000      DO 30 K=1,NPTS
007433 000      30  COST(J,K)=COST(J,K)+PI-(IPI(J)+IPI(K))
007434 000      IF(ICOST.EQ.0) GO TO 300
007435 000      IF(PRINTM)CALL SHOW(NPTS,COST)

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007436 000 IF(PRINTM) WRITE(6,2001) C
007437 000 IF(PRINTM) CALL TIMEOUT
007438 000 C* REORDER CONNECTIONS
007439 000 C* COMPUTE MINIMUM COST
007440 000 T=TREE(1,IPI,IPO)
007441 000 C=CMINP3(ISTART,J1,J2)
007442 000 C=C+T
007443 000 IF(C.LE.CMAX)GO TO 40
007444 000 C* NEW MAXIMUM FOUND
007445 000 IT=0
007446 000 CMAX=C
007447 000 CALL COPY(NPTS,COST,CTEMP)
007448 000 K1=J1
007449 000 K2=J2
007450 000 DO 35 J=1,NPTS
007451 000 ITEMP(J)=IPI(J)
007452 000 35 ITEMP(J)=IPO(J)
007453 000 GO TO 25
007454 000 40 IT=IT+1
007455 000 IF(IT.LT.MAXIT)GO TO 25
007456 000 C* GO BACK TO BEST COST
007457 000 CALL COPY(NPTS,CTEMP,COST)
007458 000 IF(IFOUT(JPI).NE.0)CALL IFOUT(JPI),1,NPTS,COST)
007459 000 WRITE(6,2001) CMAX
007460 000 CALL TIMEOUT
007461 000 IF(JPI.EQ.NPI)GO TO 900
007462 000 IT=0
007463 000 JPI=JPI+1
007464 000 PI=PISAVE(JPI)
007465 000 WRITE(6,2100) PI
007466 000 GO TO 23
007467 000 800 FOUND=.TRUE.
007468 000 900 WRITE(6,2000) CSAVE,ISTART
007469 000 WRITE(6,2001) CMAX
007470 000 2001 FORMAT(' MAXIMUM COST',F7.0)
007471 000 CALL TIMEOUT
007472 000 RETURN
007473 000 END
007474 000 DELT,SI TUMER-S*NONSYM.TIMEOUT,,,152574103711
007475 000 SUBROUTINE TIMEOUT
007476 000 K=ITIME(1,J)
007477 000 T=1/5000.
007478 000 WRITE(6,2000)T
007479 000 2000 FORMAT(' TIME',F10.3,' SEC')
007480 000 RETURN
007481 000 END
007482 000 DELT,SI TUMER-S*NONSYM.TREE,,,112152063412
007483 000 FUNCTION TREE(N,IPI,IPO)
007484 000 INCLUDE BLANK
007485 000 DIMENSION IPI(MAXPTS),IPO(MAXPTS)
007486 000 DIMENSION LINKI(MAXPTS),LINKO(MAXPTS),CLINK(MAXPTS)
007487 000 M=NPTS-N-1
007488 000 C=0.
007489 000 IF(M.LE.0) GO TO 900
007490 000 C* FIND A ROW NOT IN INLIST AND INITIALIZE
007491 000 IROW=0
007492 000 DO 10 I=1,NPTS

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007490	000	IPI(I)=-1
007494	000	IPO(I)=-1
007495	000	LINKO(I)=0
007496	000	IF(.NOT.IN(I)) IROW=1
007497	000	LINKI(I)=0
007498	000	C* FIND LEAST CONNECTION TO IROW
007499	000	C=1.0E38
007500	000	DO 40 J=1,NPTS
007501	000	IF(IN(J).OR.J.EQ.IROW) GO TO 30
007502	000	IF(COST(IROW,J).GT.COST(I,IROW)) GO TO 30
007503	000	LINKI(J)=IROW
007504	000	CLINK(J)=COST(IROW,J)
007505	000	GO TO 35
007506	000	30 LINKO(J)=IROW
007507	000	CLINK(J)=COST(J,IROW)
007508	000	35 IF(CLINK(J).GE.C) GO TO 40
007509	000	ICOL=J
007510	000	C=CLINK(J)
007511	000	40 CONTINUE
007512	000	C* SET NEW VECTORS FOR NEXT BRANCH
007513	000	IF(CLINK(ICOL).EQ.0) GO TO 45
007514	000	IPI(ICOL)=IPI(ICOL)+1
007515	000	IROW=LINKI(ICOL)
007516	000	IPO(IROW)=IPO(IROW)+1
007517	000	LINKI(ICOL)=0
007518	000	GO TO 50
007519	000	45 IPO(ICOL)=IPO(ICOL)+1
007520	000	IROW=LINKO(ICOL)
007521	000	IPI(IROW)=IPI(IROW)+1
007522	000	LINKO(ICOL)=0
007523	000	50 MEM=1
007524	000	IF(M.EQ.0) GO TO 900
007525	000	C* ADD OTHER BRANCHES TO MEM
007526	000	DO 80 J=1,M
007527	000	CMIN=1.0E38
007528	000	DO 70 I=1,NPTS
007529	000	MES=LINKI(I)+LINKO(I)
007530	000	IF(MES.EQ.0) GO TO 75
007531	000	IF(CLINK(I).LE.COST(I,ICOL)) GO TO 60
007532	000	LINKI(I)=0
007533	000	LINKO(I)=ICOL
007534	000	CLINK(I)=COST(I,ICOL)
007535	000	60 IF(CLINK(I).LE.COST(ICOL,I)) GO TO 65
007536	000	LINKI(I)=ICOL
007537	000	LINKO(I)=0
007538	000	CLINK(I)=COST(ICOL,I)
007539	000	65 IF(CLINK(I).GE.CMIN) GO TO 70
007540	000	JCOL=I
007541	000	CMIN=CLINK(I)
007542	000	70 CONTINUE
007543	000	ICOL=JCOL
007544	000	IF(LINKI(ICOL).EQ.0) GO TO 75
007545	000	IPI(ICOL)=IPI(ICOL)+1
007546	000	IROW=LINKI(ICOL)
007547	000	IPO(IROW)=IPO(IROW)+1
007548	000	LINKI(ICOL)=0
007549	000	GO TO 80

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007550      000      75      IPO(ICOL)=IPO(ICOL)+1
007551      000          IROW=LINKO(ICOL)
007552      000          IPI(IROW)=IPI(IROW)+1
007553      000          LINKO(ICOL)=0
007554      000      80      C=C+CLINK(ICOL)
007555      000      999      TREE=C
007556      000          RETURN
007557      000          END
007558      000      @PDP,FI TUMER-S*NONSYM.COMMON/BLANK,,156514061612
007559      000      BLANK      PROC
007560      000          PARAMETER MAXPTS=60,MXLYS=60
007561      000          COMMON NPTS,NLIST,CMAX,CMX,NMAXPTS),LIST(2,MXLYS),
007562      000          *ISAVE(MAXPTS),COST(MAXPTS,NMAXPTS),EX(MAXPTS,MAXPTS),
007563      000          *CMAX,FOUND
007564      000          LOGICAL IN,EX,FOUND
007565      000          END

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END ELT.  
@FIN

RUNID: IERON ACCT: 011A0936 PROJECT: PARTN

IERON\*MSG: CHG AND EXCESS LINES TO E-24-200

TIME: TOTAL: 00:00:33.017

CPU: 00:00:09.476 I/O: 00:00:01.002

CC/FR: 00:00:21.639 WAIT: 00:00:00.000

IMAGES READ: 7569 PAGES: 133

START: 16:58:34 FEB 21,1975 FIN: 17:01:36 FEB 21,1975